

Capstone Project - Predicting Patient No-Shows Using Appointment Data

```
library(tidyverse)
library(lubridate)
library(caret)
library(randomForest)
library(GGally)
```

Read data and assign to *appointments*

```
appointments <- read_csv("Final_Data.csv")
```

```
## Parsed with column specification:
## cols(
##   kept_status = col_character(),
##   appt_date = col_character(),
##   appt_time = col_time(format = ""),
##   appt_length = col_integer(),
##   date_scheduled = col_character(),
##   patient_age = col_integer(),
##   patient_gender = col_character(),
##   billing_type = col_character(),
##   prior_missed = col_integer(),
##   prior_kept = col_integer(),
##   patient_distance = col_integer(),
##   office_zip = col_character(),
##   provider_specialty = col_character(),
##   remind_call_result = col_character()
## )
```

```
zipcodes <- read_csv("zipcodes.csv")
```

```
## Parsed with column specification:
## cols(
##   office_zip = col_character(),
##   county_code = col_character(),
##   city_size = col_integer()
## )
```

Data Summary and Structure

```
summary(appointments)
```

```
## kept_status      appt_date      appt_time      appt_length
## Length:342862    Length:342862    Length:342862    Min.   : 10
## Class :character  Class :character  Class1:hms       1st Qu.: 60
## Mode  :character  Mode  :character  Class2:difftime  Median : 60
##                                     Mode   :numeric   Mean   : 57
##                                     3rd Qu.: 60
##                                     Max.   :600
##
```

```
## date_scheduled      patient_age      patient_gender      billing_type
## Length:342862      Min.   : 0.00      Length:342862      Length:342862
## Class :character    1st Qu.: 17.00      Class :character    Class :character
## Mode :character     Median : 34.00      Mode :character     Mode :character
##                      Mean    : 35.56
##                      3rd Qu.: 54.00
##                      Max.    :264.00
##
## prior_missed         prior_kept         patient_distance      office_zip
## Min.   : 0.000      Min.   : 0.00      Min.   : 0.0      Length:342862
## 1st Qu.: 1.000      1st Qu.: 2.00      1st Qu.: 0.0      Class :character
## Median : 2.000      Median : 6.00      Median : 3.0      Mode :character
## Mean   : 2.451      Mean   : 8.02      Mean   : 10.8
## 3rd Qu.: 3.000      3rd Qu.: 11.00     3rd Qu.: 9.0
## Max.   :117.000     Max.   :676.00     Max.   :2688.0
##                      NA's    :974
## provider_specialty remind_call_result
## Length:342862      Length:342862
## Class :character    Class :character
## Mode :character     Mode :character
##
##
##
##
```

```
str(appointments)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':   342862 obs. of  14 variables:
## $ kept_status      : chr  "Kept" "Kept" "Kept" "Kept" ...
## $ appt_date        : chr  "9/1/16" "9/1/16" "9/1/16" "9/1/16" ...
## $ appt_time         :Classes 'hms', 'difftime' atomic [1:342862] 19800 28800 28800 28800 28800 28800 ...
## ..- attr(*, "units")= chr "secs"
## $ appt_length       : int   90 60 120 60 60 60 60 60 60 90 ...
## $ date_scheduled    : chr  "8/1/16" "1/18/16" "2/3/16" "6/8/16" ...
## $ patient_age       : int    7 75 31 45 49 71 49 38 36 13 ...
## $ patient_gender    : chr   "Male" "Female" "Male" "Male" ...
## $ billing_type      : chr   "DMAP" "Commercial" "DMAP" "DMAP" ...
## $ prior_missed      : int    1 2 1 6 5 6 8 0 2 3 ...
## $ prior_kept        : int    3 5 5 15 6 6 20 0 5 12 ...
## $ patient_distance   : int   41 29 5 5 0 5 0 539 0 4 ...
## $ office_zip        : chr   "AP" "BL" "BL" "BL" ...
## $ provider_specialty: chr   "A" "A" "A" "B" ...
## $ remind_call_result: chr   "Left Message" "Answered - Confirmed" "Left Message" "Answered - No Resp
## - attr(*, "spec")=List of 2
## ..$ cols      :List of 14
## .. ..$ kept_status      : list()
## .. ..- attr(*, "class")= chr  "collector_character" "collector"
## .. ..$ appt_date        : list()
## .. ..- attr(*, "class")= chr  "collector_character" "collector"
## .. ..$ appt_time        :List of 1
## .. .. ..$ format: chr ""
## .. ..- attr(*, "class")= chr  "collector_time" "collector"
## .. ..$ appt_length      : list()
## .. ..- attr(*, "class")= chr  "collector_integer" "collector"
## .. ..$ date_scheduled   : list()
```

```
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## .. ..$ patient_age : list()
## .. ..- attr(*, "class")= chr "collector_integer" "collector"
## .. ..$ patient_gender : list()
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## .. ..$ billing_type : list()
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## .. ..$ prior_missed : list()
## .. ..- attr(*, "class")= chr "collector_integer" "collector"
## .. ..$ prior_kept : list()
## .. ..- attr(*, "class")= chr "collector_integer" "collector"
## .. ..$ patient_distance : list()
## .. ..- attr(*, "class")= chr "collector_integer" "collector"
## .. ..$ office_zip : list()
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## .. ..$ provider_specialty: list()
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## .. ..$ remind_call_result: list()
## .. ..- attr(*, "class")= chr "collector_character" "collector"
## ..$ default: list()
## .. ..- attr(*, "class")= chr "collector_guess" "collector"
## ..- attr(*, "class")= chr "col_spec"
```

```
head(appointments[,1:5])
```

```
## # A tibble: 6 x 5
##   kept_status appt_date appt_time appt_length date_scheduled
##   <chr>      <chr>      <time>      <int> <chr>
## 1 Kept      9/1/16      05:30         90 8/1/16
## 2 Kept      9/1/16      08:00         60 1/18/16
## 3 Kept      9/1/16      08:00        120 2/3/16
## 4 Kept      9/1/16      08:00         60 6/8/16
## 5 Missed    9/1/16      08:00         60 6/28/16
## 6 Kept      9/1/16      08:00         60 7/12/16
```

```
head(appointments[,6:10])
```

```
## # A tibble: 6 x 5
##   patient_age patient_gender billing_type prior_missed prior_kept
##   <int> <chr>      <chr>      <int>      <int>
## 1      7 Male      DMAP         1         3
## 2     75 Female    Commercial    2         5
## 3     31 Male      DMAP         1         5
## 4     45 Male      DMAP         6        15
## 5     49 Male      Commercial    5         6
## 6     71 Male      DMAP         6         6
```

```
head(appointments[,11:14])
```

```
## # A tibble: 6 x 4
##   patient_distance office_zip provider_specialty remind_call_result
##   <int> <chr>      <chr>      <chr>
## 1      41 AP      A      Left Message
## 2      29 BL      A      Answered - Confirmed
## 3       5 BL      A      Left Message
## 4       5 BL      B      Answered - No Response
```

```
## 5          0 BL          B          Answered - No Response
## 6          5 BL          A          Answered - Confirmed
```

```
# Check for NAs
sapply(appointments, function(x) sum(is.na(x)))
```

```
##      kept_status      appt_date      appt_time
##           0           0           0
##      appt_length    date_scheduled    patient_age
##           0           0           0
##      patient_gender    billing_type    prior_missed
##           0           0           0
##      prior_kept    patient_distance    office_zip
##           0           974           0
## provider_specialty remind_call_result
##           0           0
```

patient_distance variable has 972 NA values.

Data Dictionary

```
variable_description <- c(
  "Dependent variable: kept or missed",
  "Appointment date",
  "Appointment time",
  "Appointment length in minutes",
  "Date appointment was scheduled",
  "Patient age",
  "Patient gender",
  "Billing type",
  "Number of prior missed appointments",
  "Number of prior kept appointments",
  "Patient distance from office in miles",
  "Office Zip Code - Anonymized",
  "Provider primary specialty code",
  "Reminder Call result")
variable <- colnames(appointments)

as_data_frame(cbind(c(1:length(variable)),variable, variable_description))
```

```
## # A tibble: 14 x 3
##      V1      variable      variable_description
##    <chr> <chr>      <chr>
##  1 1      kept_status    Dependent variable: kept or missed
##  2 2      appt_date      Appointment date
##  3 3      appt_time      Appointment time
##  4 4      appt_length    Appointment length in minutes
##  5 5      date_scheduled  Date appointment was scheduled
##  6 6      patient_age     Patient age
##  7 7      patient_gender   Patient gender
##  8 8      billing_type    Billing type
##  9 9      prior_missed    Number of prior missed appointments
## 10 10     prior_kept        Number of prior kept appointments
## 11 11     patient_distance  Patient distance from office in miles
```

```
## 12 12    office_zip      Office Zip Code - Anonymized
## 13 13    provider_specialty Provider primary specialty code
## 14 14    remind_call_result Reminder Call result
```

Will combine the appointment time and date into one variable, `appt_datetime`.

```
appointments_2 <- appointments %>%
  mutate(appt_datetime = lubridate::mdy_hms(paste(appt_date, appt_time)))
appointments_2$date_scheduled <- as.POSIXct(appointments_2$date_scheduled,
  format = "%m/%d/%y")
```

Calculating percent of missed appointments overall. Will first create a logical variable *missed*, where 1 represents a missed appointment and 0 represents a kept appointment.

```
appointments_2 <- appointments_2 %>%
  mutate(missed = ifelse(appointments_2$kept_status == "Missed", 1,0))
missed_rate <- mean(appointments_2$missed)
missed_rate
```

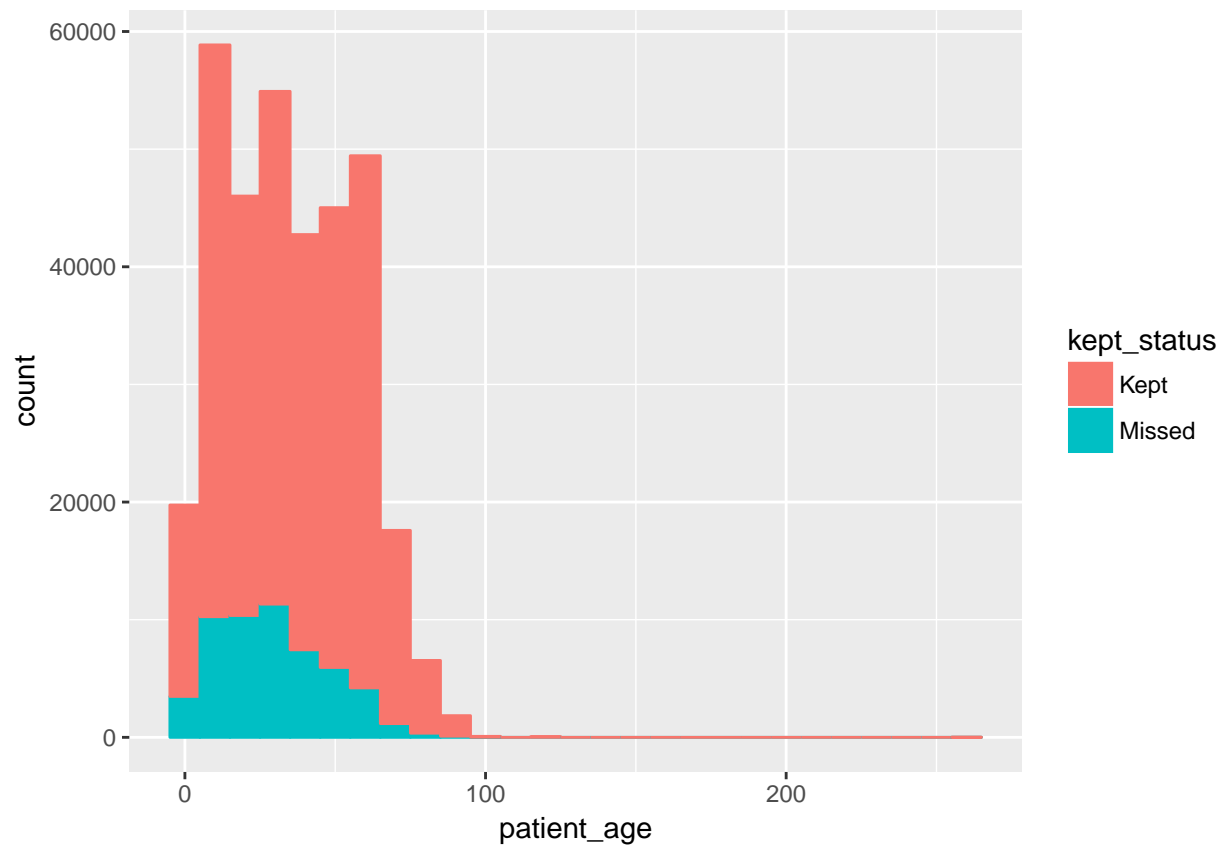
```
## [1] 0.1592944
```

About 16% of the total appointments are missed.

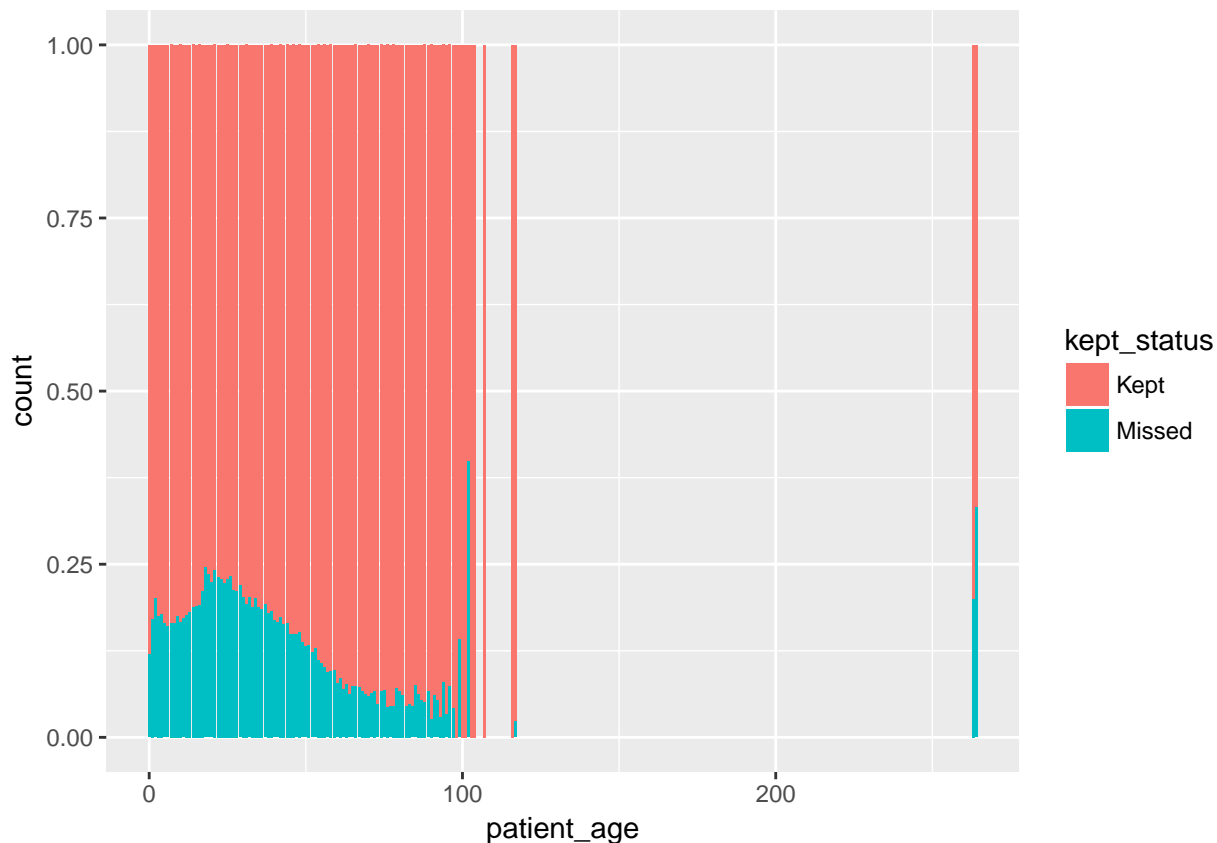
Data Exploration

`patient_age`

```
ggplot(
  data = appointments_2,
  mapping = aes(x = patient_age)
) +
  geom_histogram(
    mapping = aes(x = patient_age, col = kept_status, fill = kept_status),
    binwidth = 10)
```



```
ggplot(data = appointments_2) +  
  geom_bar(  
    mapping = aes(x = patient_age, fill = kept_status),  
    position = "fill")
```



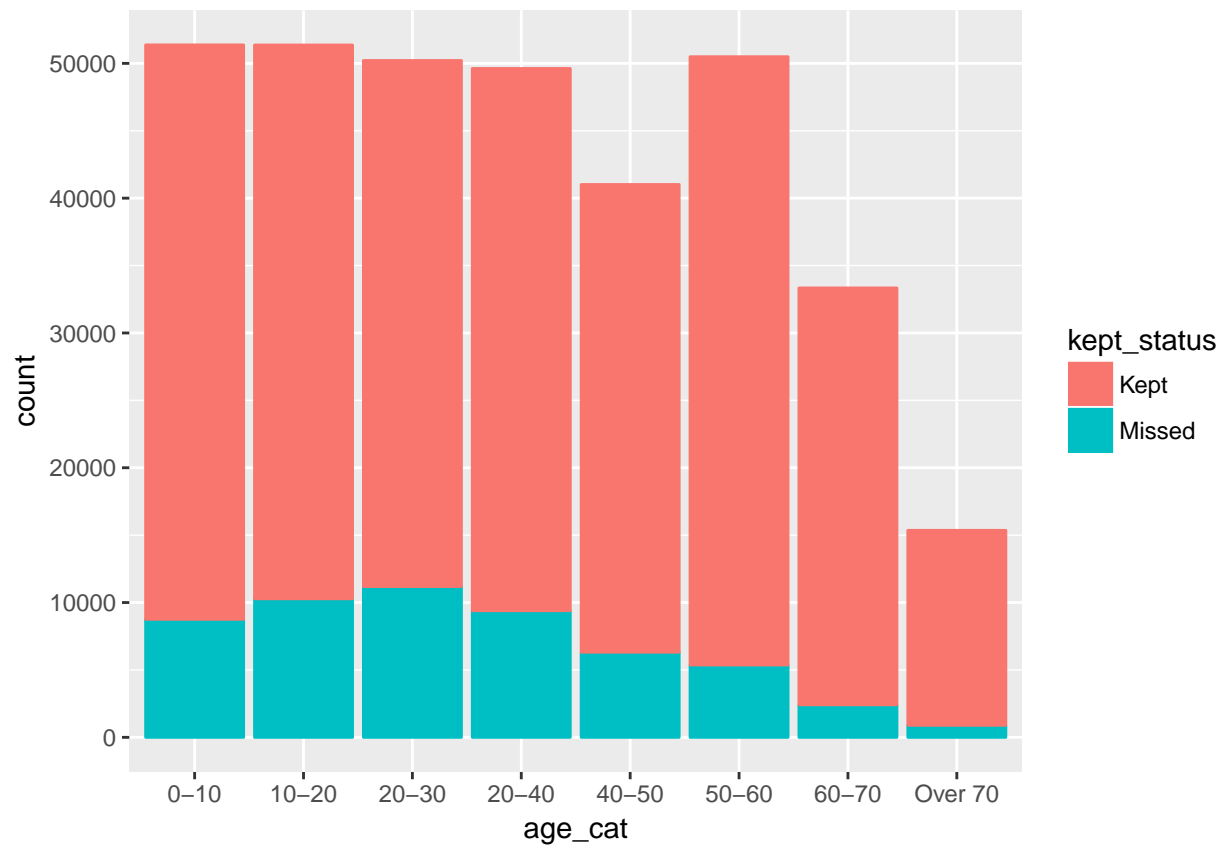
Ranges from 0-264, so there are obviously a few impossible values. Ratio of missed appointments decreases with age in general.

Removing observations of ages greater than 100, creating categorical age groups and replotting.

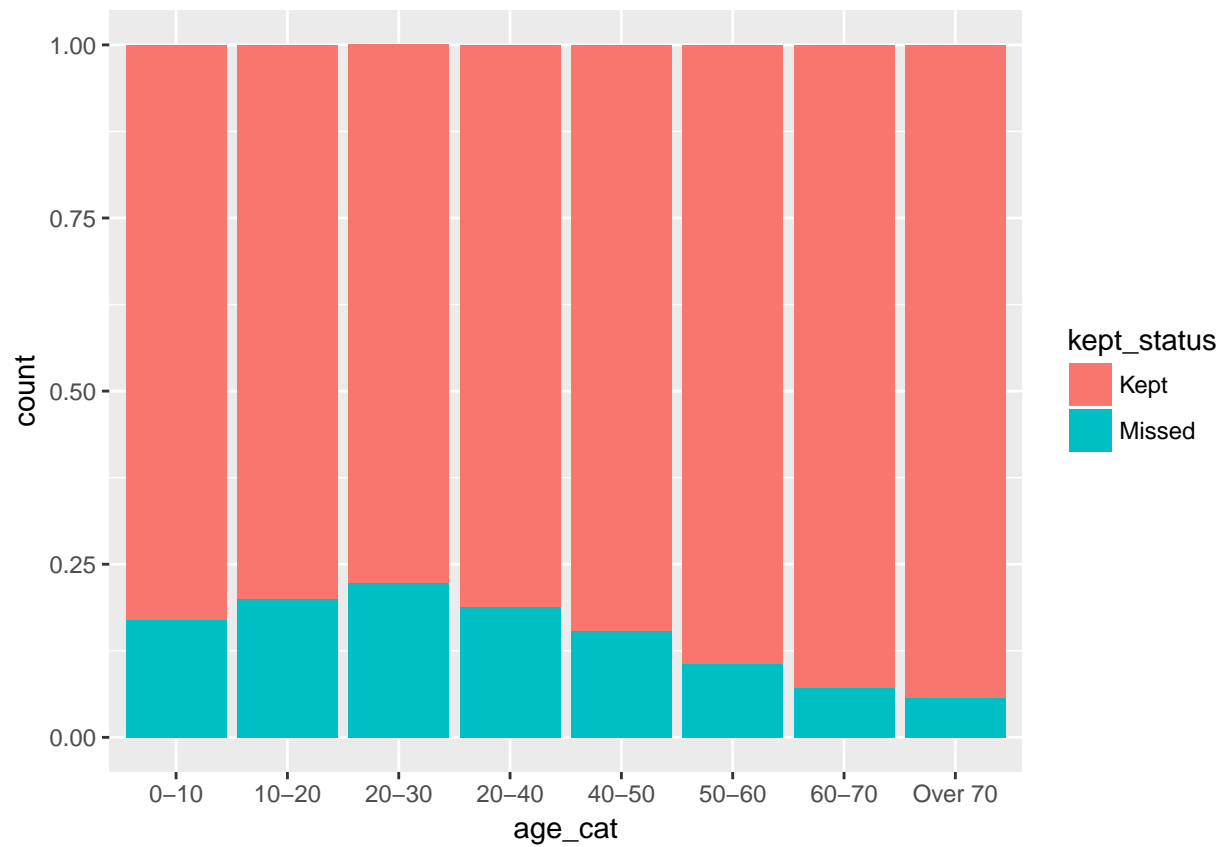
```
appointments_2 <- appointments_2 %>%
  filter(patient_age <= 100)

appointments_2 <- appointments_2 %>%
  mutate(age_cat = cut(patient_age, breaks = c(-1, 10, 20, 30, 40, 50, 60, 70, 101),
    labels = c("0-10", "10-20", "20-30", "20-40", "40-50",
      "50-60", "60-70", "Over 70")))

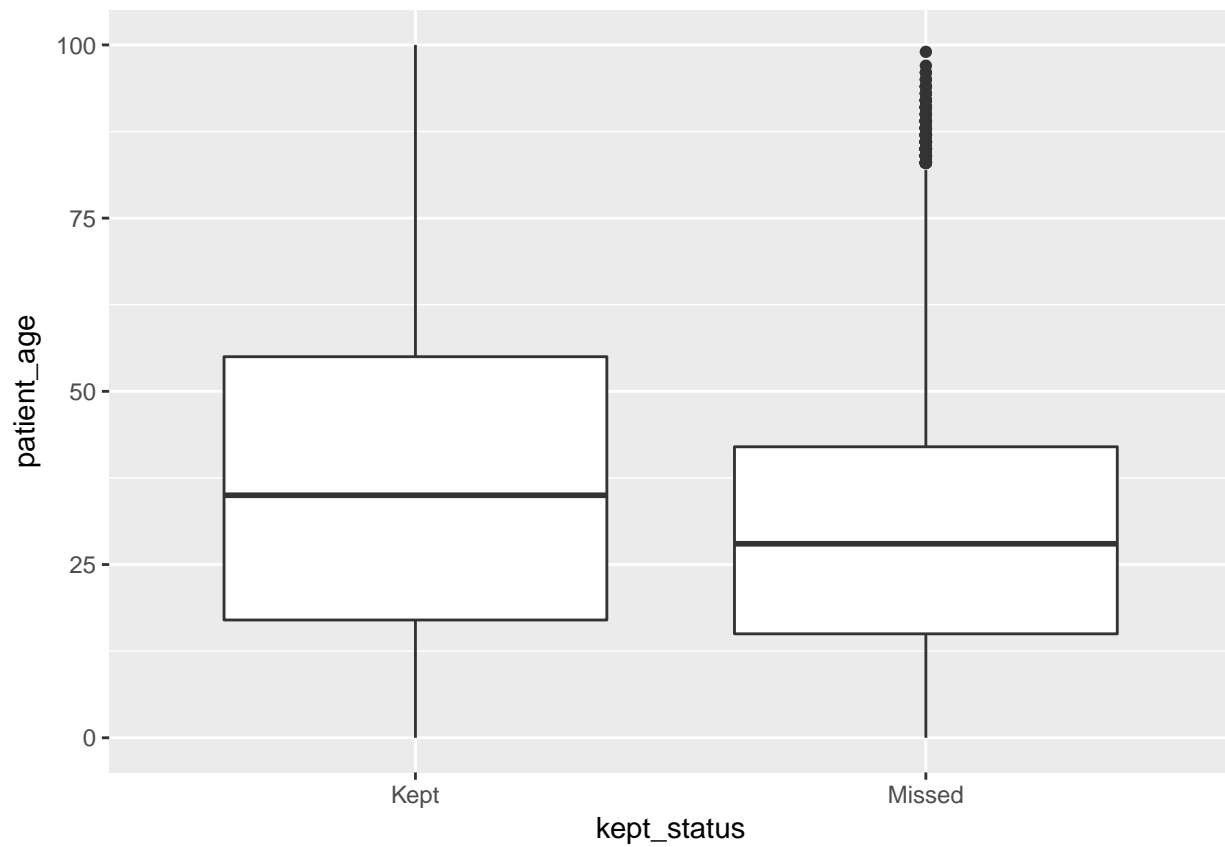
ggplot(appointments_2, aes(x = age_cat, group = kept_status, col = kept_status,
  fill = kept_status)) +
  stat_count()
```



```
ggplot(data = appointments_2) +  
  geom_bar(mapping = aes(x = age_cat, fill = kept_status), position = "fill")
```

```
ggplot(data = appointments_2, aes(x = kept_status, y = patient_age)) +  
  geom_boxplot()
```

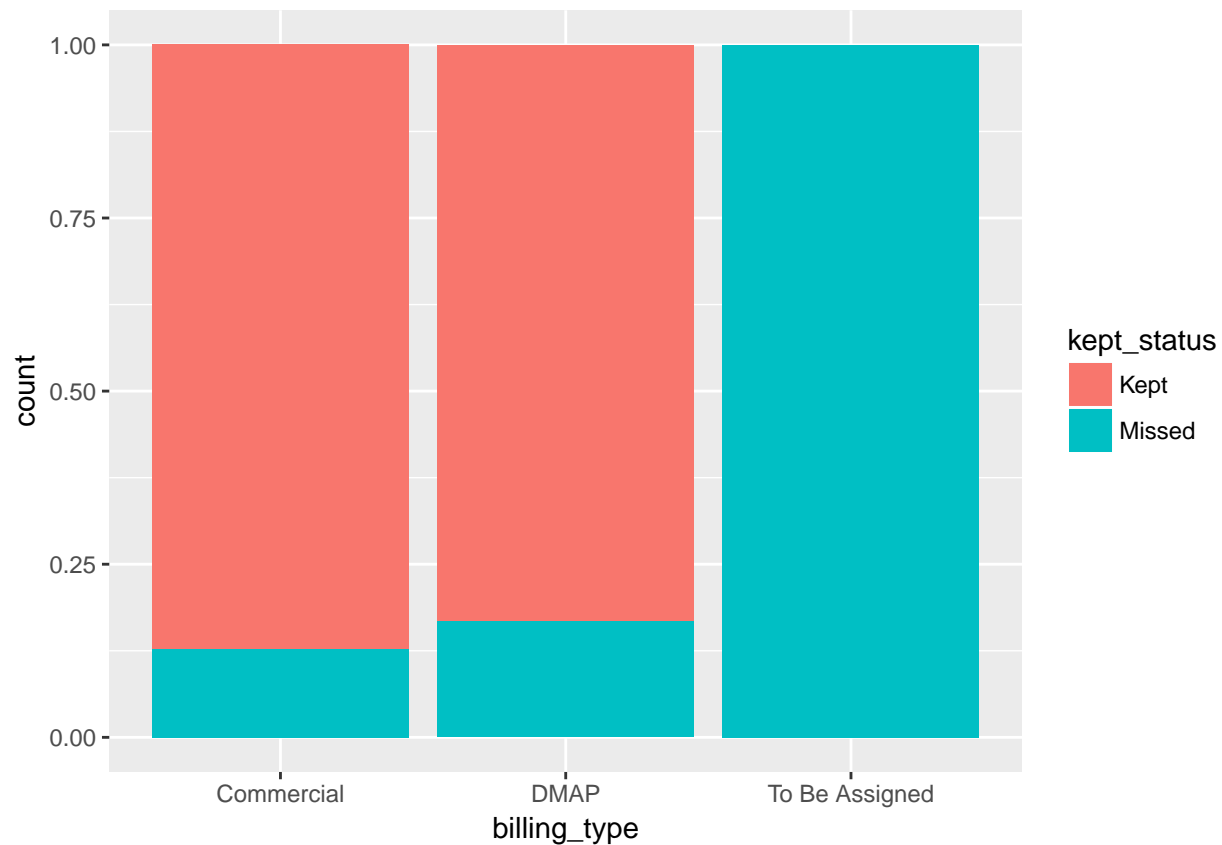


billing_type

```
table(appointments_2$billing_type)
```

```
##
##      Commercial      DMAP To Be Assigned
##      78278         264486             1
```

```
ggplot(data = appointments_2) +
  geom_bar(mapping = aes(x = billing_type, fill = kept_status), position = "fill")
```



Only one row has *To Be Assigned* value and will just be removed There is a minor difference between billing types. DMAP has a higher proportion of missed appointments

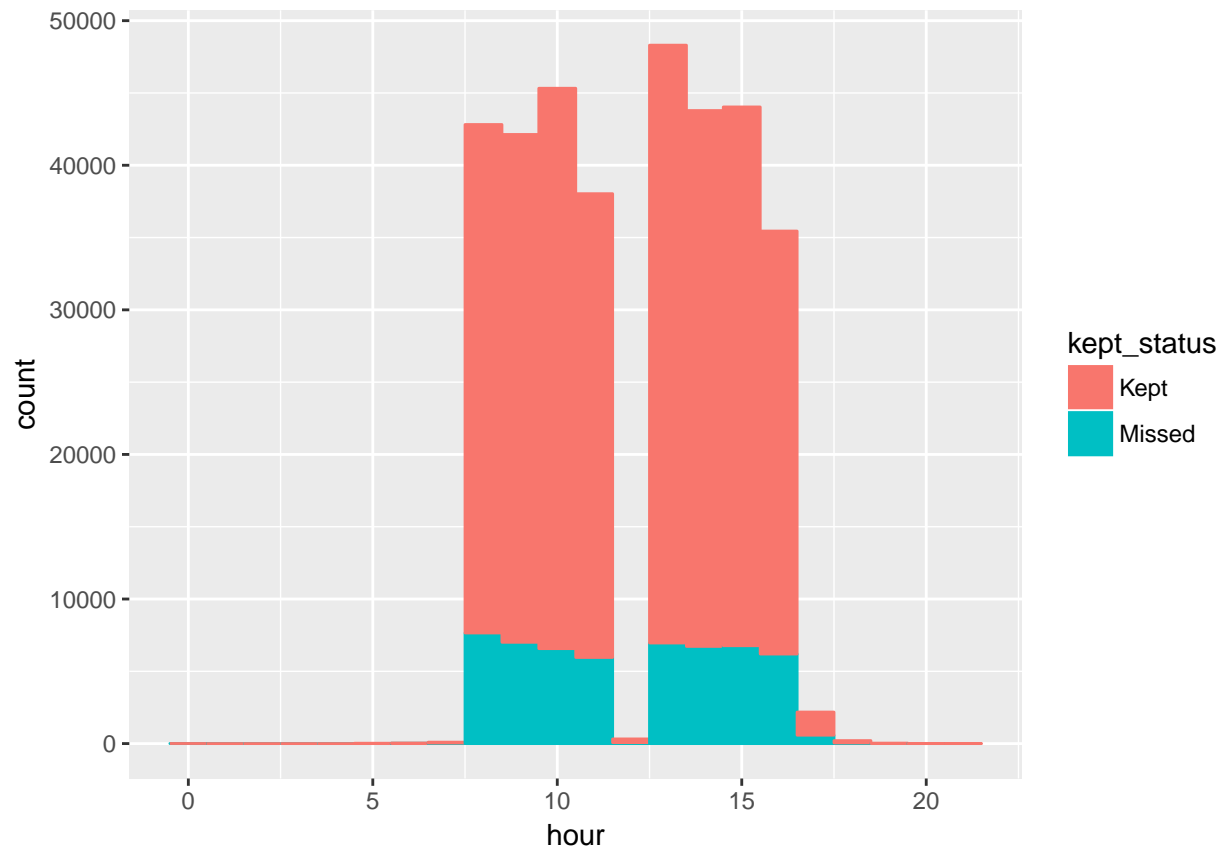
```
appointments_2 <- subset(appointments_2,
                          appointments_2$billing_type != "To Be Assigned")
```

appt_datetime

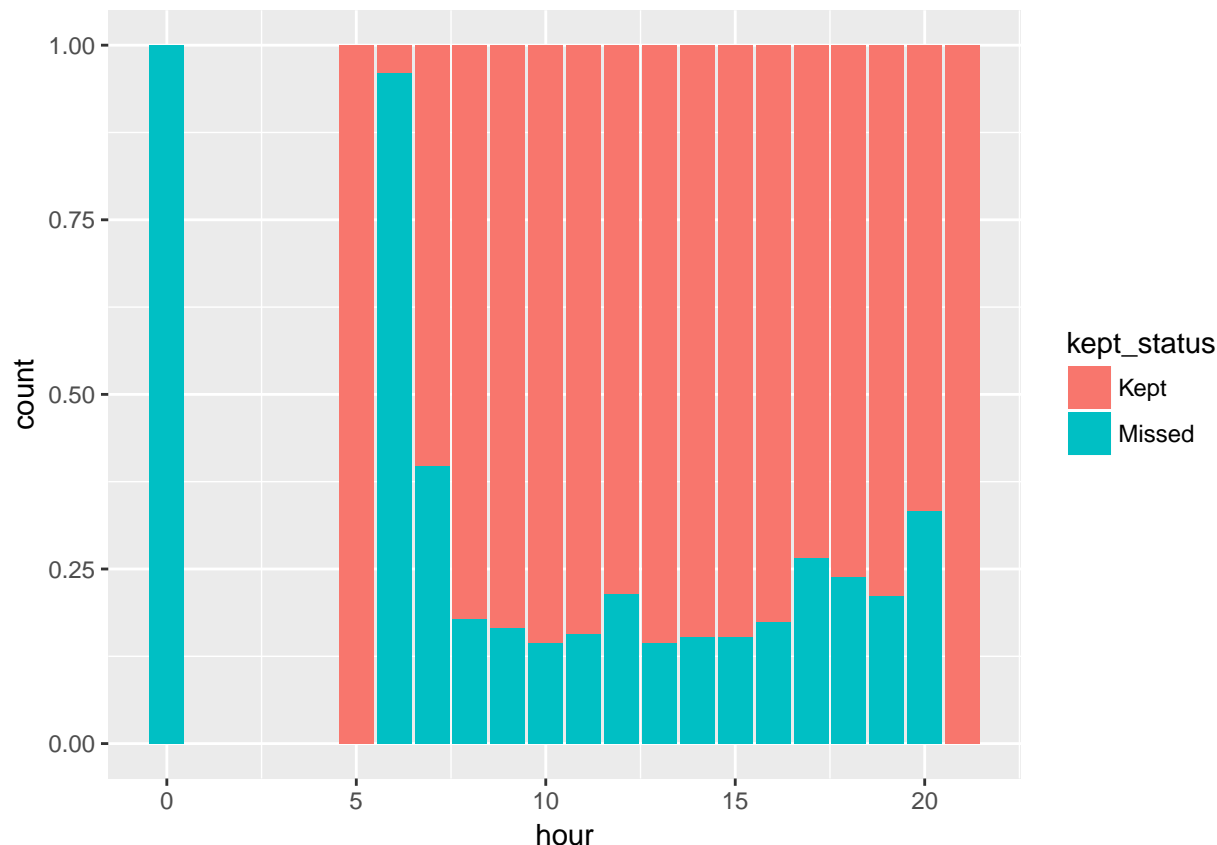
Creating new *hour* variable and plot by hour

```
appointments_2 <- appointments_2 %>%
  mutate(hour = lubridate::hour(appointments_2$appt_datetime))

ggplot(data = appointments_2,
       aes(x = hour, group = kept_status, col = kept_status, fill = kept_status)) +
  geom_histogram(binwidth = 1)
```



```
ggplot(data = appointments_2) +  
  geom_bar(mapping = aes(x = hour, fill = kept_status), position = "fill")
```



Ranges from 00:00:00 to 21:00:00. More appointments are missed in the early morning, late afternoon and early evening, and around lunchtime, however, there are very few appointments at these times. During main scheduling periods, the variation is less significant.

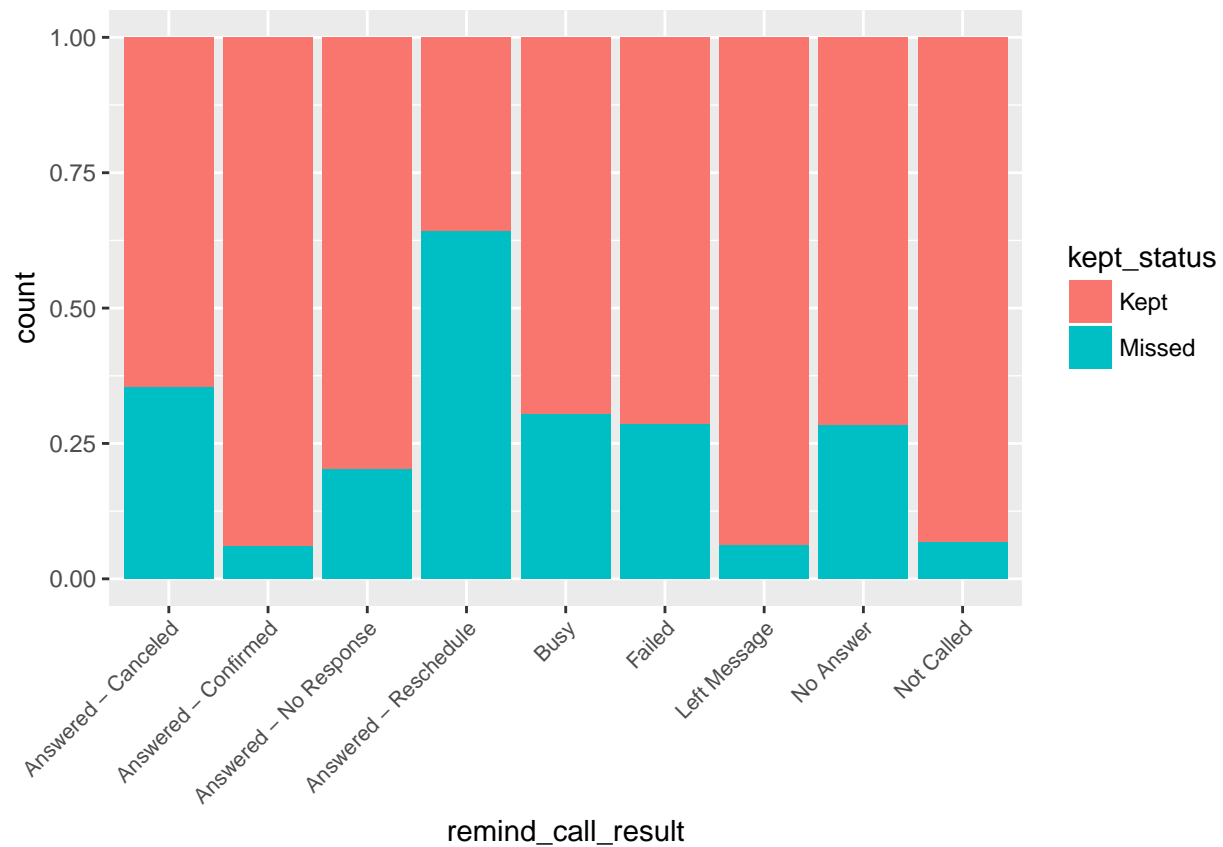
remind_call_result

```
table(appointments_2$remind_call_result)
```

```
##
##      Answered - Canceled      Answered - Confirmed Answered - No Response
##              152              49108              180860
## Answered - Reschedule              Busy              Failed
##              1369              1104              27943
##           Left Message              No Answer              Not Called
##              18429              377              63422
```

Low counts of “Answered - Cancelled”, “Answered - Reschedule”, “Busy”, and “No Answer”

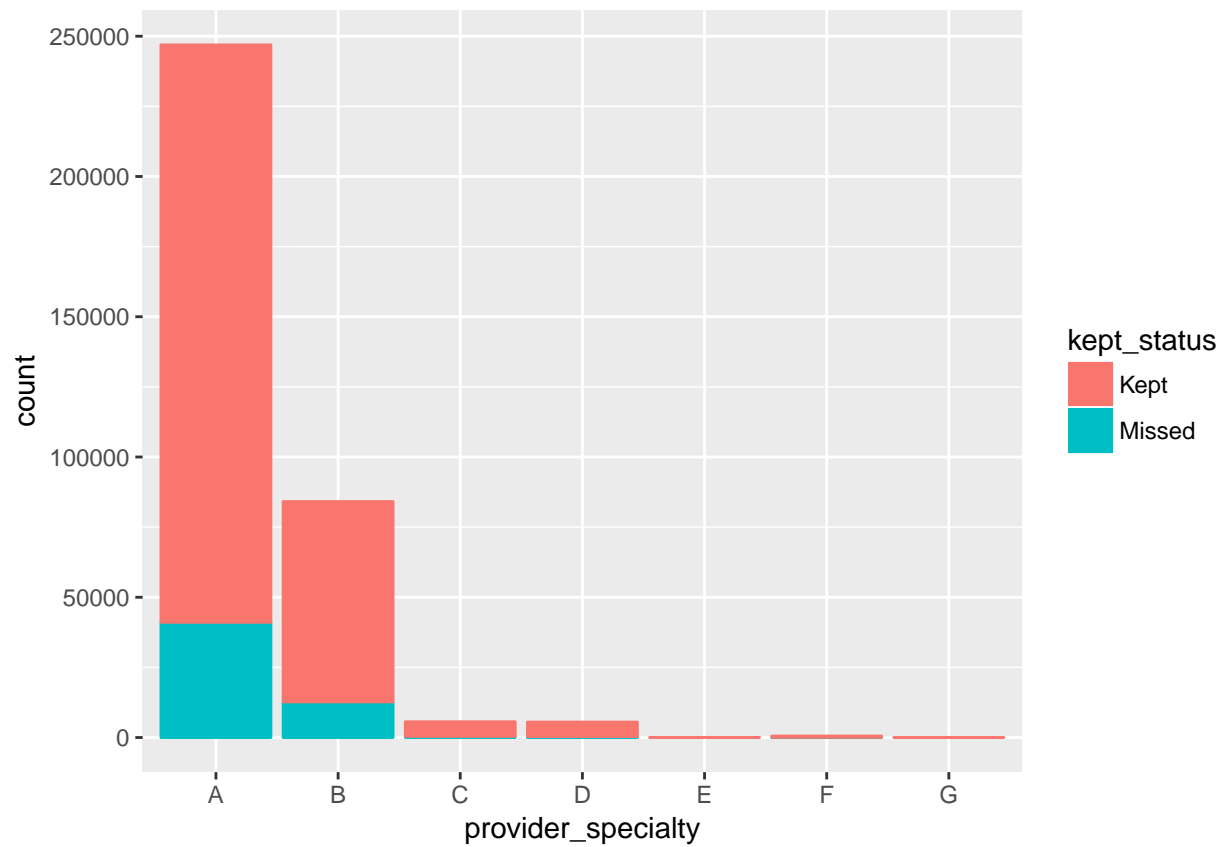
```
ggplot(data = appointments_2) +
  geom_bar(
    mapping = aes(x = remind_call_result, fill = kept_status),
    position = "fill"
  ) +
  theme(
    axis.text.x = element_text(size = 8, angle = 45,
    hjust = 1, vjust = 1))
```



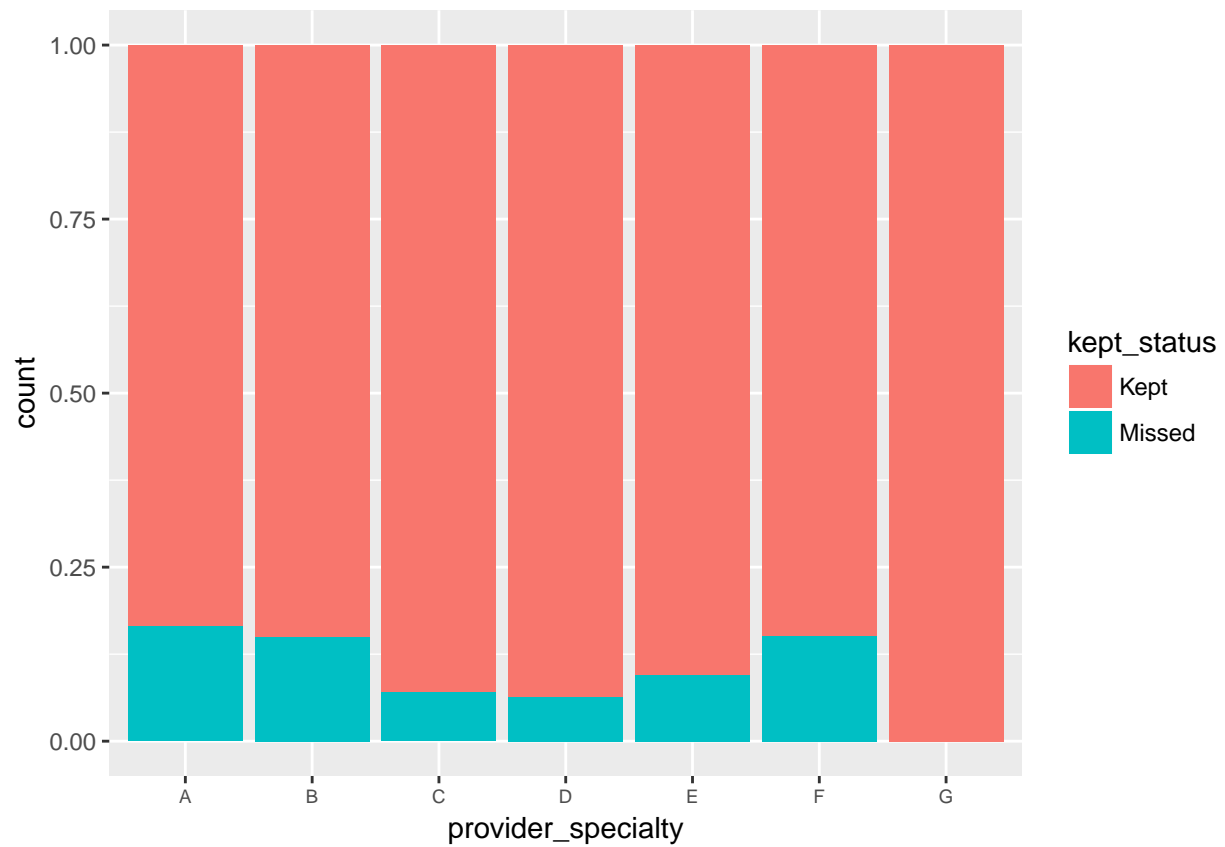
~65% of appointments with “Answered - Cancelled” and ~35% with “Answered-Reschedule” still kept their appointments, however, very few observations in these categories.

provider_specialty

```
ggplot(
  data = appointments_2,
  mapping = aes(x = provider_specialty, col = kept_status, fill = kept_status)
) +
  stat_count()
```



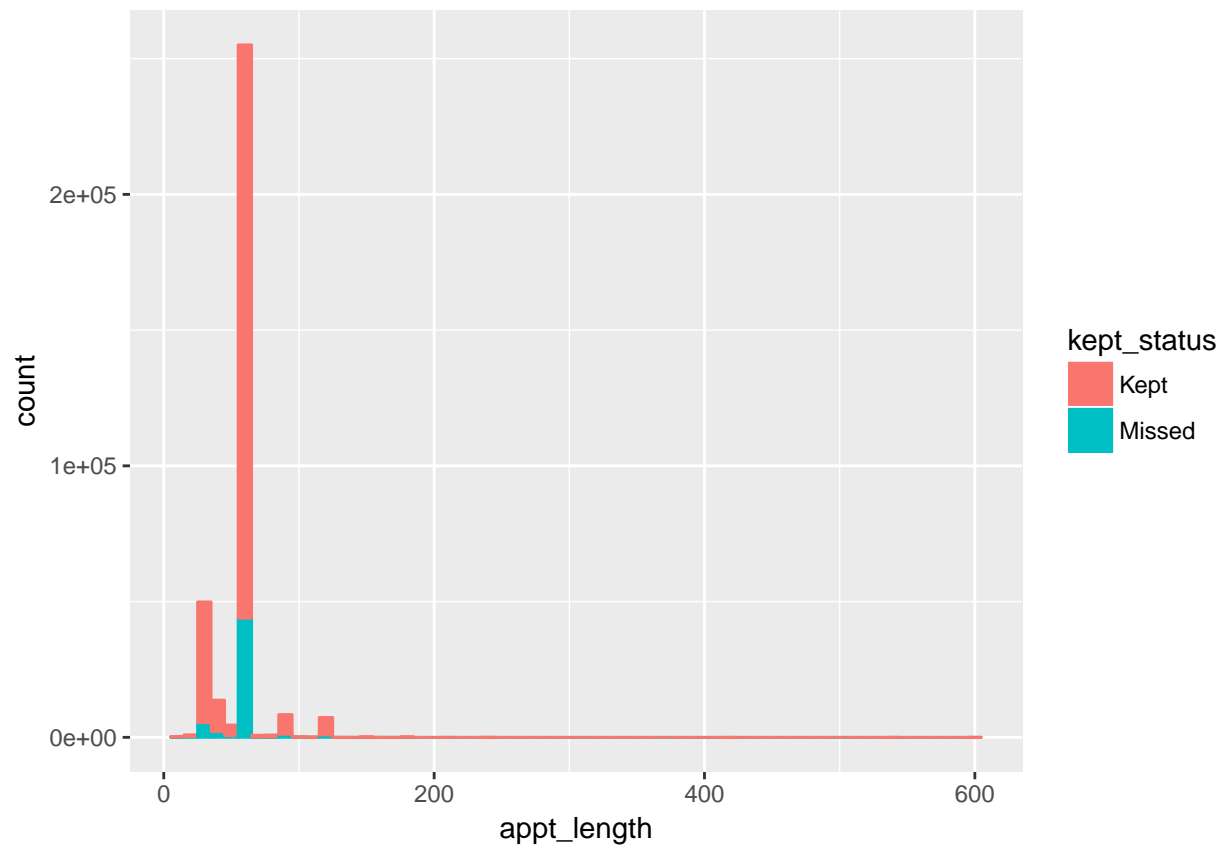
```
ggplot(data = appointments_2) +  
  geom_bar(  
    mapping = aes(x = provider_specialty, fill = kept_status),  
    position = "fill"  
  ) +  
  theme(axis.text.x = element_text(size = 7))
```



C, D, and E provider specialties have lower proportion of missed appointments,

appt_length

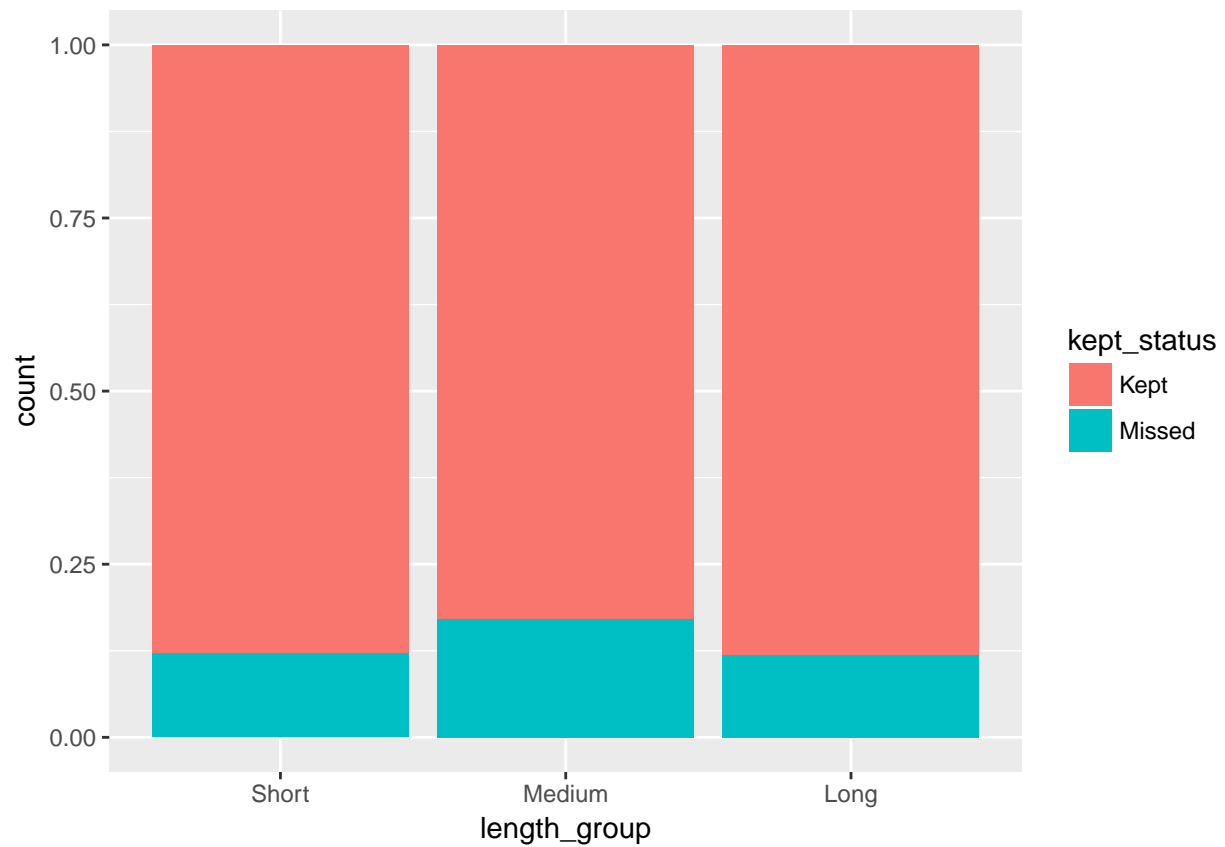
```
ggplot(  
  data = appointments_2,  
  mapping = aes(x = appt_length, col = kept_status, fill = kept_status)  
) +  
  geom_histogram(binwidth = 10)
```

Most Appointments appear to be 60 minutes long. 30-minute appointments are next most popular.

```
appointments_2 <- appointments_2 %>%
  mutate(length_group = cut(appt_length, breaks = c(-1, 45, 75, 1000), labels = c("Short", "Medium", "Long")))

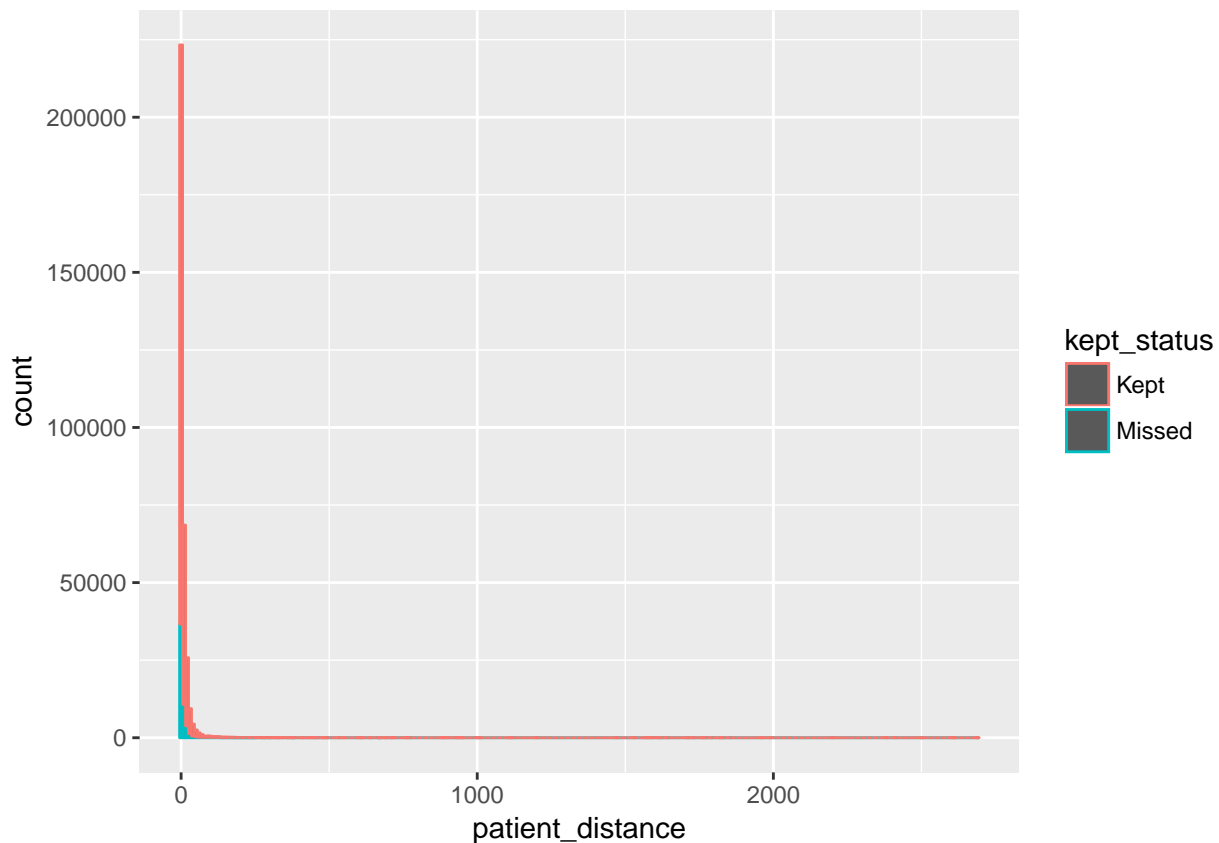
ggplot(data = appointments_2) +
  geom_bar(mapping = aes(x = length_group, fill = kept_status), position = "fill")
```



patient_distance

```
ggplot(appointments_2, aes(x = patient_distance, group = kept_status, col = kept_status)) +  
  geom_histogram(binwidth = 10)
```

Warning: Removed 972 rows containing non-finite values (stat_bin).



patient_distance is very right-skewed, therefore NA values will be replaced with median rather than mean.

```
appointments_2$patient_distance <- appointments_2$patient_distance %>%
  tidyr::replace_na(median(appointments_2$patient_distance, na.rm = TRUE))
```

Create new variables

percent_missed = percent of prior appointments missed. New represents represents first time appointments
 appt_lead_time is the difference between the day the appointment was scheduled and the day of the appointment.

```
appointments_3 <- appointments_2 %>%
  mutate(percent_missed = prior_missed / (prior_missed + prior_kept)) %>%
  mutate(new = ifelse(appointments_2$prior_missed == 0 & appointments_2$prior_kept == 0, 1, 0)) %>%
  mutate(appt_lead_time = date(appt_datetime) - date(date_scheduled)) %>%
  mutate(weekday = strftime(appt_datetime, "%A"))
```

Add city_size and county_code from zipcode data.

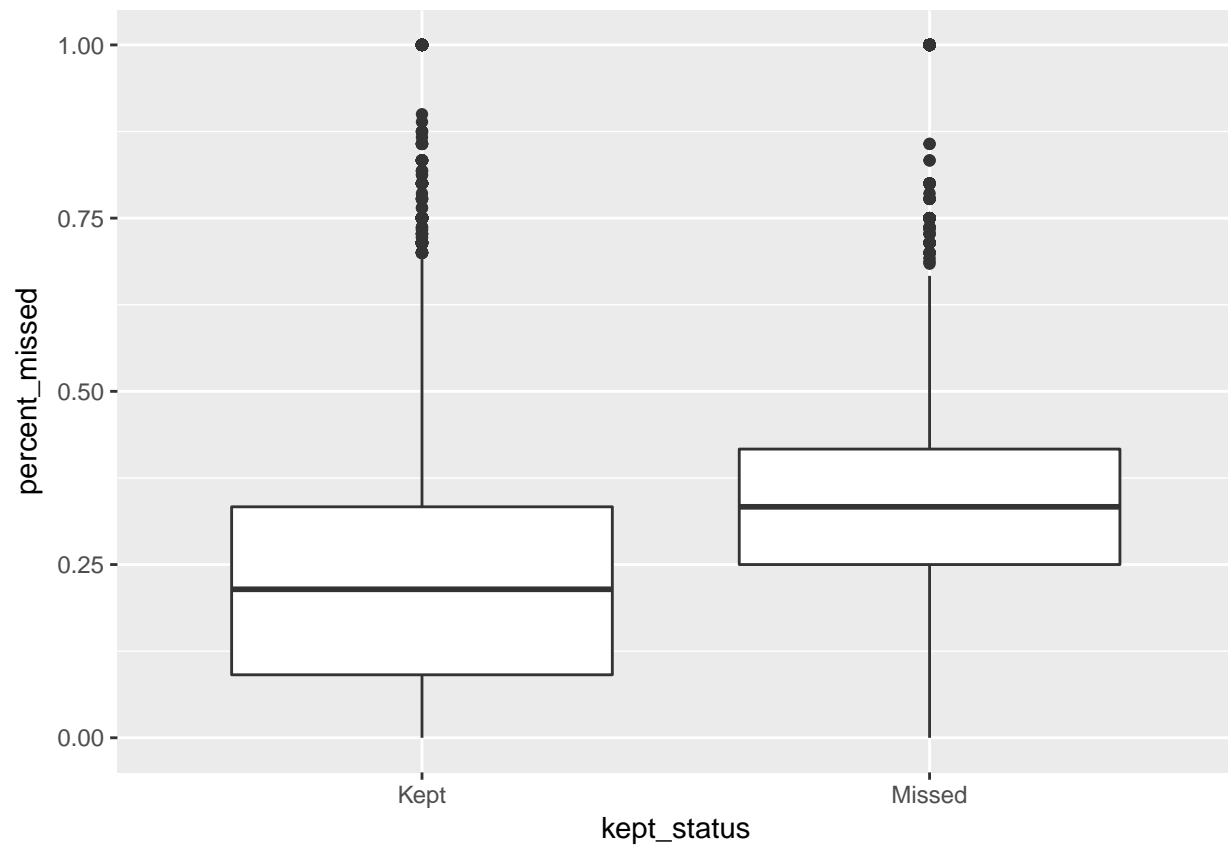
```
appointments_3 <- dplyr::left_join(appointments_3, zipcodes, by = "office_zip")
```

percent_missed

Create random subset and plot

```
ggplot(data = appointments_3, aes(x = kept_status, y = percent_missed)) +
  geom_boxplot()
```

```
## Warning: Removed 22338 rows containing non-finite values (stat_boxplot).
```



new

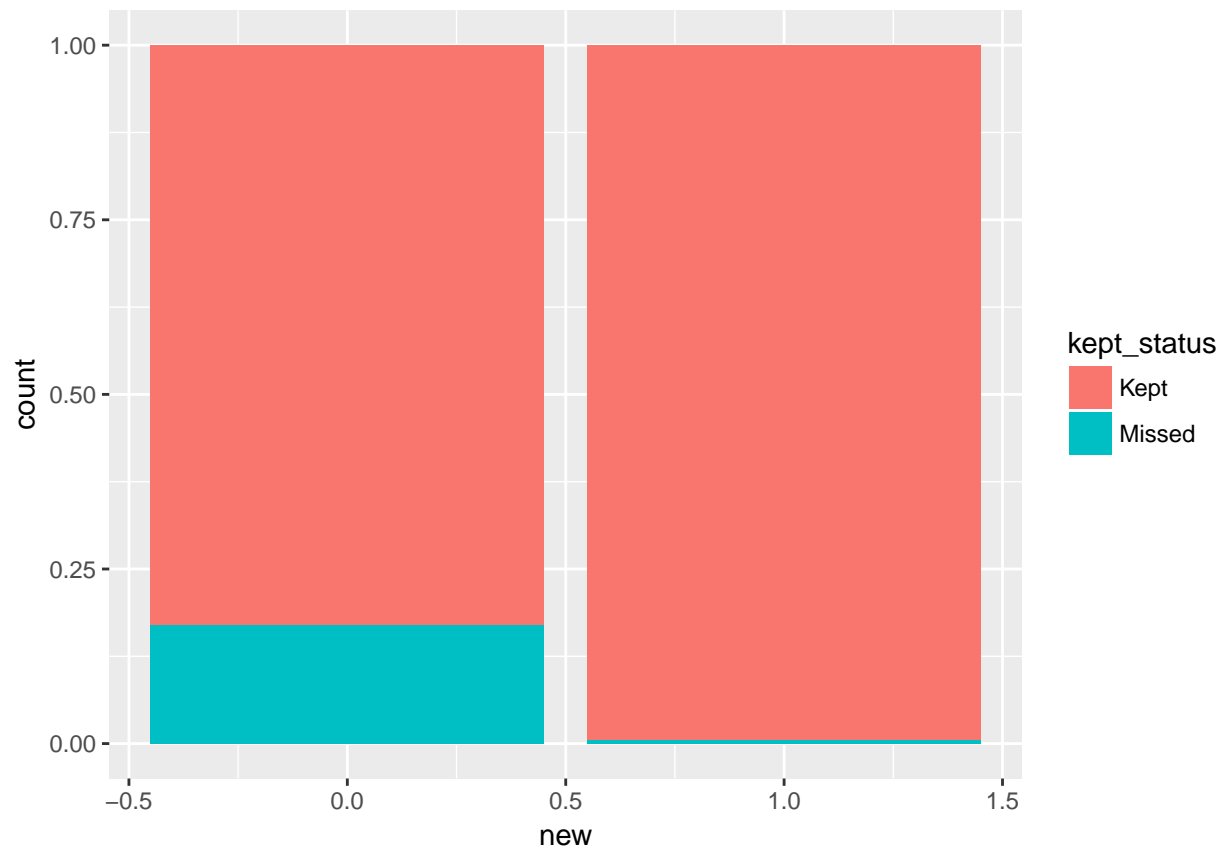
```
table(appointments_3$new)
```

```
##
```

```
##      0      1
```

```
## 320426 22338
```

```
ggplot(data = appointments_3) +  
  geom_bar(mapping = aes(x = new, fill = kept_status), position = "fill")
```

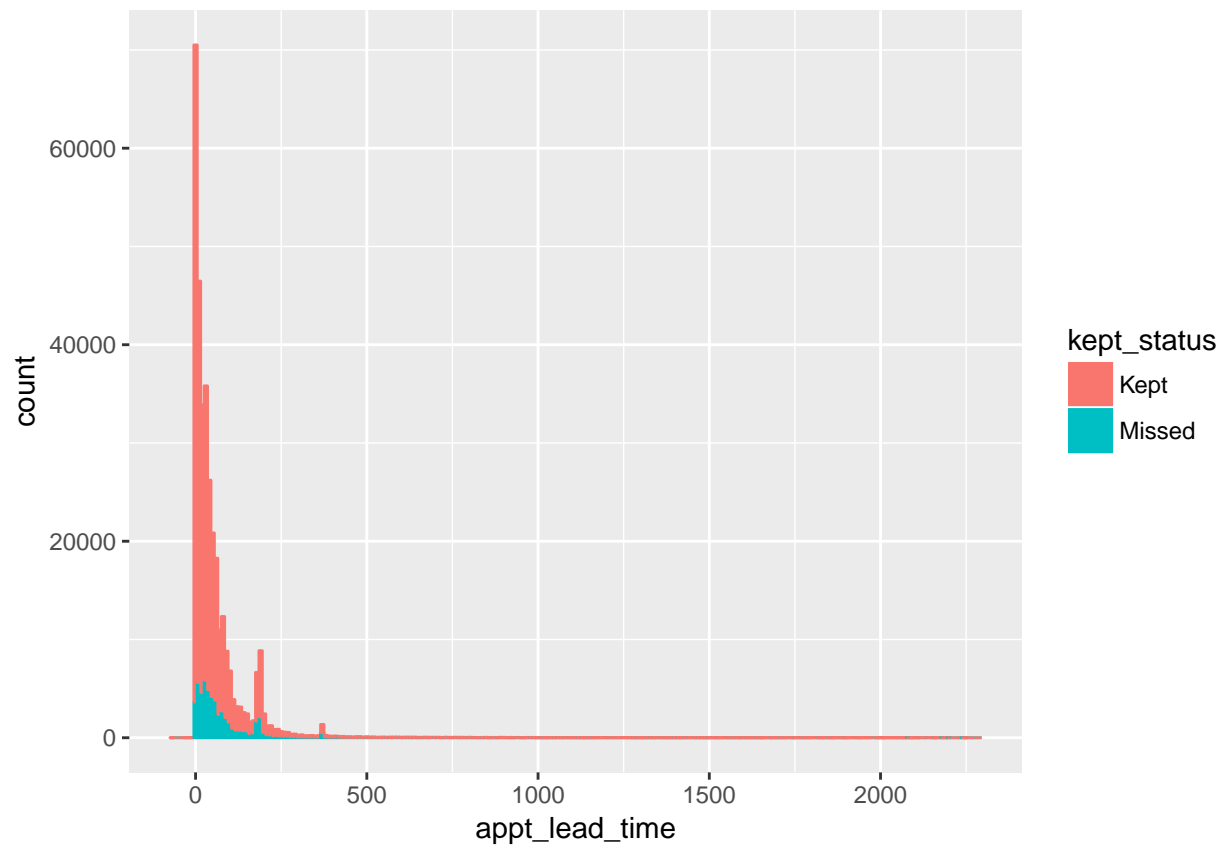


New patients have a very high percentage of kept appointments. 22k of 342k appointments are first-time, or about 6.4%

appt_lead_time

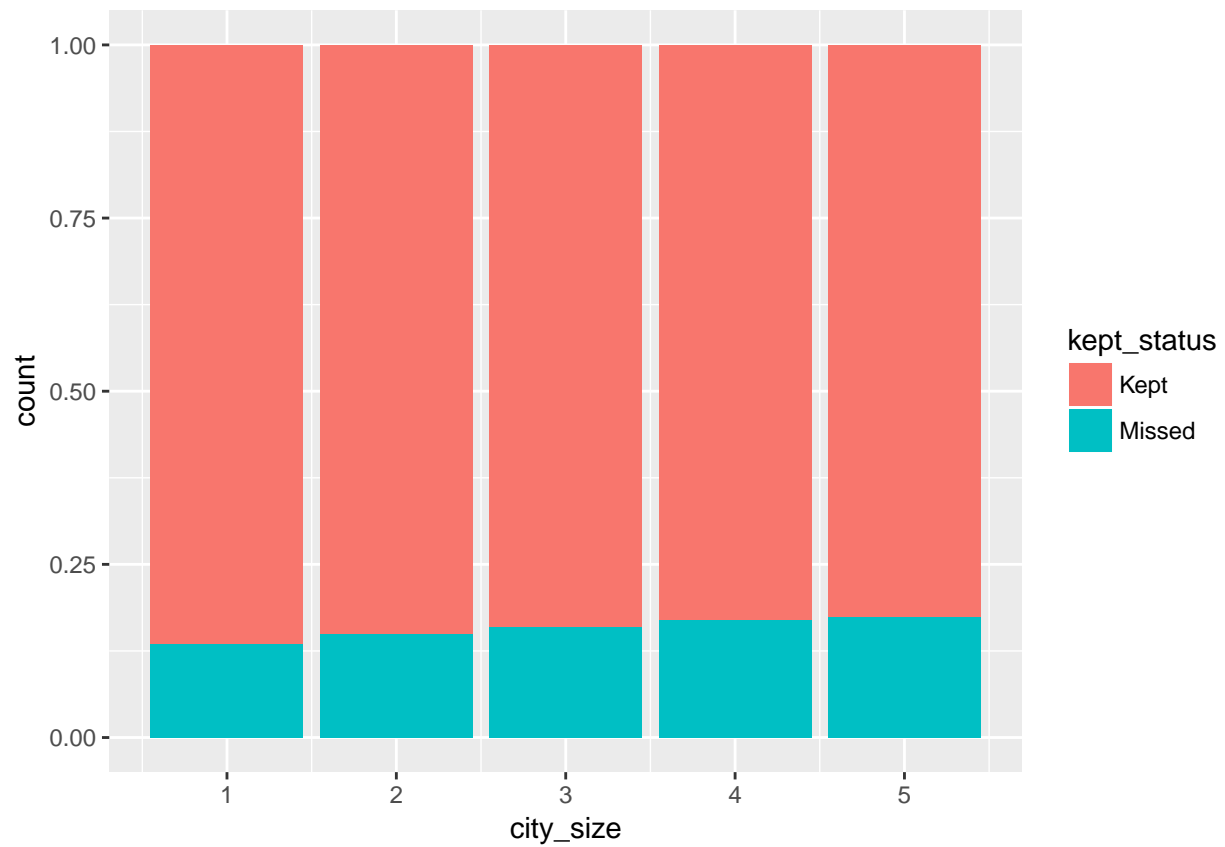
```
ggplot(appointments_3, aes(x = appt_lead_time, col = kept_status, fill = kept_status)) +  
  geom_histogram(binwidth = 10)
```

Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.



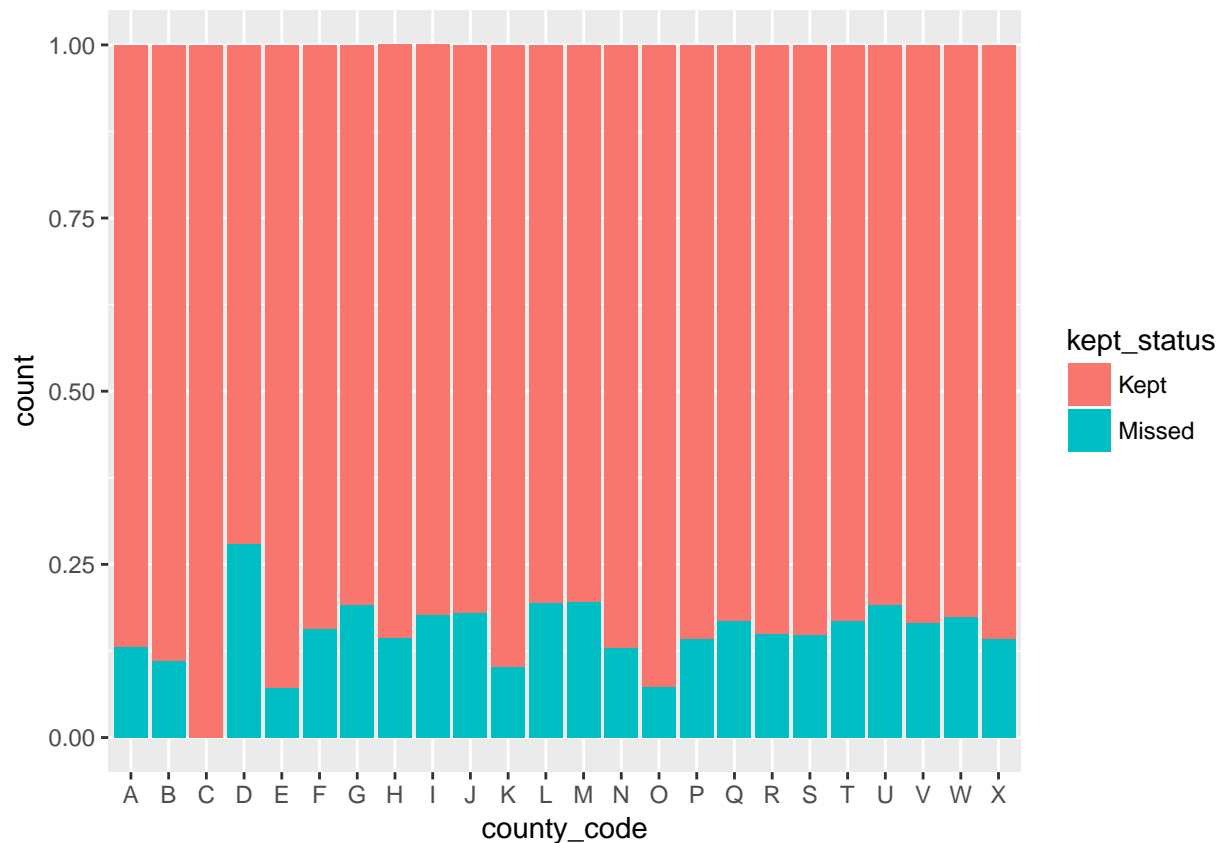
city_size

```
ggplot(data = appointments_3) +  
  geom_bar(mapping = aes(x = city_size, fill = kept_status), position = "fill")
```



county_code

```
ggplot(data = appointments_3) +  
  geom_bar(mapping = aes(x = county_code, fill = kept_status), position = "fill")
```



```
#appointments_sample_025 <- appointments_3 %>%
#   sample_frac(size = 0.025, replace = FALSE)
#ggpairs(data = appointments_sample_05[,20:24], cardinality_threshold = 50)
```

Modeling

Create Modeling Data

```
model_data <- appointments_3 %>%
model_data$new <- as.factor(model_data$new)
model_data$percent_missed <- as.integer(model_data$percent_missed * 100)
#Replace NAs with mean
model_data$percent_missed <- model_data$percent_missed %>%
  tidyr::replace_na(mean(model_data$percent_missed, na.rm = TRUE))
factor_columns <- c("kept_status", "patient_gender", "billing_type",
"office_zip", "provider_specialty", "remind_call_result", "hour", "weekday",
"county_code", "length_group")
model_data[factor_columns] <- lapply(model_data[factor_columns], factor)
#Check for NAs
sapply(model_data, function(x) sum(is.na(x)))
```

```
##      kept_status      appt_date      appt_time
##           0           0           0
##      appt_length    date_scheduled    patient_age
##           0           0           0
##      patient_gender    billing_type    prior_missed
##           0           0           0
```



```
##      prior_kept    patient_distance      office_zip
##           0           0           0
## provider_specialty remind_call_result    appt_datetime
##           0           0           0
##           missed           age_cat           hour
##           0           0           0
##      length_group    percent_missed           new
##           0           0           0
##      appt_lead_time           weekday    county_code
##           0           0           0
##      city_size
##           0
```

```
str(model_data)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':  342764 obs. of  25 variables:
## $ kept_status      : Factor w/ 2 levels "Kept","Missed": 1 1 1 1 2 1 1 1 1 1 ...
## $ appt_date        : chr  "9/1/16" "9/1/16" "9/1/16" "9/1/16" ...
## $ appt_time        :Classes 'hms', 'difftime'  atomic [1:342764] 19800 28800 28800 28800 28800 28800 28800 28800 ...
## .. ..- attr(*, "units")= chr "secs"
## $ appt_length      : int   90 60 120 60 60 60 60 60 60 90 ...
## $ date_scheduled   : POSIXct, format: "2016-08-01" "2016-01-18" ...
## $ patient_age      : int    7 75 31 45 49 71 49 38 36 13 ...
## $ patient_gender    : Factor w/ 4 levels "Female","Male",...: 2 1 2 2 2 2 2 1 2 2 ...
## $ billing_type     : Factor w/ 2 levels "Commercial","DMAP": 2 1 2 2 1 2 1 1 2 2 ...
## $ prior_missed     : int    1 2 1 6 5 6 8 0 2 3 ...
## $ prior_kept       : int    3 5 5 15 6 6 20 0 5 12 ...
## $ patient_distance : num   41 29 5 5 0 5 0 539 0 4 ...
## $ office_zip       : Factor w/ 50 levels "AA","AB","AC",...: 16 38 38 38 38 38 38 38 45 34 ...
## $ provider_specialty: Factor w/ 7 levels "A","B","C","D",...: 1 1 1 2 2 1 1 1 2 1 ...
## $ remind_call_result: Factor w/ 9 levels "Answered - Canceled",...: 7 2 7 3 3 2 7 9 9 3 ...
## $ appt_datetime    : POSIXct, format: "2016-09-01 05:30:00" "2016-09-01 08:00:00" ...
## $ missed           : num    0 0 0 0 1 0 0 0 0 0 ...
## $ age_cat          : Factor w/ 8 levels "0-10","10-20",...: 1 8 4 5 5 8 5 4 4 2 ...
## $ hour             : Factor w/ 18 levels "0","5","6","7",...: 2 5 5 5 5 5 5 5 5 5 ...
## $ length_group     : Factor w/ 3 levels "Short","Medium",...: 3 2 3 2 2 2 2 2 2 3 ...
## $ percent_missed   : num   25 28 16 28 45 ...
## $ new              : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 2 1 1 ...
## $ appt_lead_time   : Class 'difftime'  atomic [1:342764] 31 227 211 85 65 51 49 0 2 664 ...
## .. ..- attr(*, "units")= chr "days"
## $ weekday          : Factor w/ 6 levels "Friday","Monday",...: 6 4 4 4 4 4 4 4 4 4 ...
## $ county_code       : Factor w/ 24 levels "A","B","C","D",...: 16 9 9 9 9 9 9 9 20 12 ...
## $ city_size        : int    2 4 4 4 4 4 4 4 4 4 ...
```

Divide model_data into train, validate, and test sets

```
train <- model_data[1:205660,]
validate <- model_data[205660:274200,]
test <- model_data[274201:nrow(model_data),]
```

```
table(train$kept_status)
```

```
##
##   Kept Missed
## 174601  31059
```

```

train2 <- train[168738:205660,]
table(train2$keptstatus)

## Warning: Unknown or uninitialised column: 'keptstatus'.
## < table of extent 0 >

train_kept <- train2[train2$kept_status == "Kept",]
train_missed <- train[train$kept_status == "Missed",]

train_balanced <- rbind(train_kept, train_missed)
table(train_balanced$kept_status)

##
##   Kept Missed
## 31059 31059

```

Logistic Regression Model

```

modell1 <- caret::train(kept_status ~ age + remindresult + specialty + billtype + hour + percent_missed +
modell1$finalModel
confusionMatrix(modell1)
##p_glm <- predict(glm, train)
#caret::confusionMatrix(p_glm, train$kept_status)

```

Random Forest Model

Using randomForest Package

```

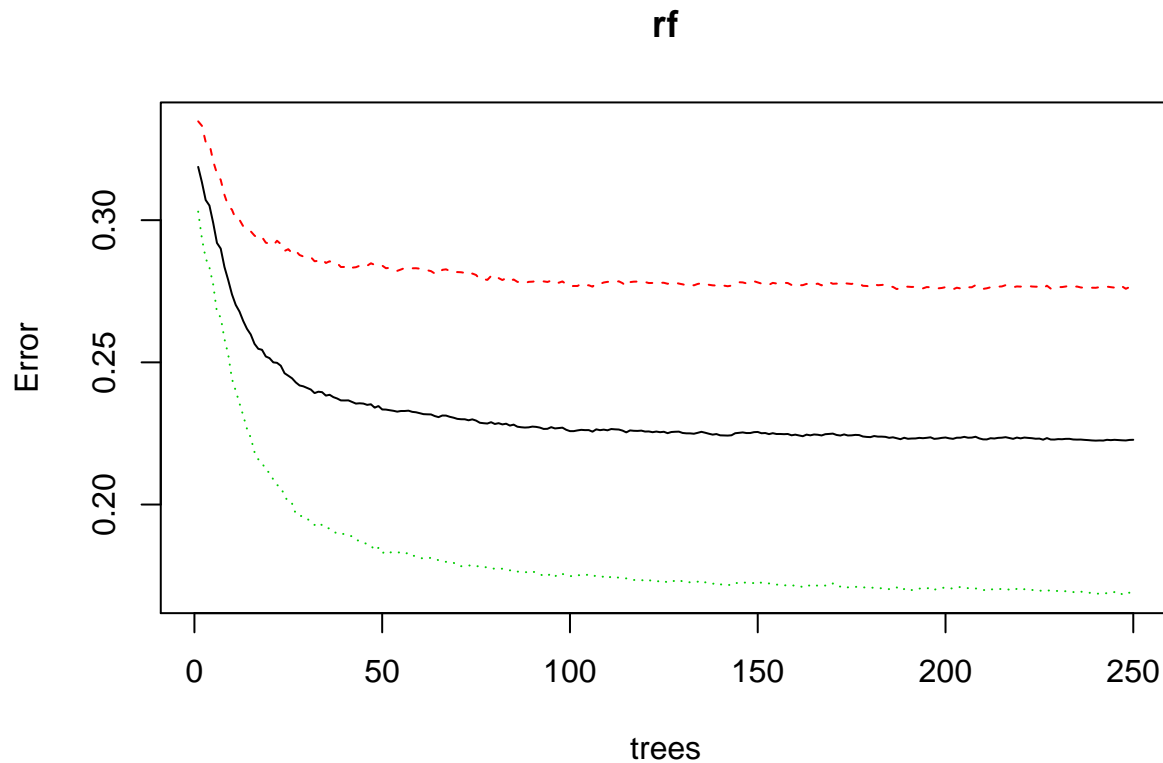
rf <- randomForest(kept_status ~ patient_age + remind_call_result + provider_specialty + billing_type +
#Takes about 30 seconds to run

print(rf)

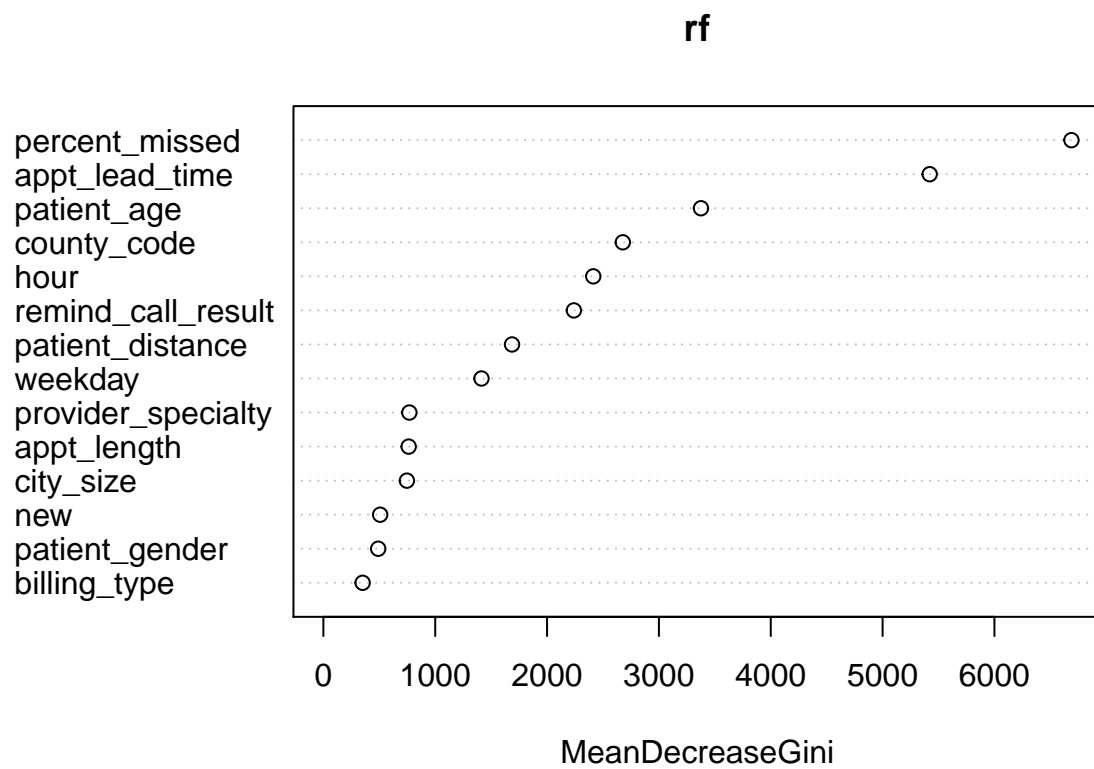
##
## Call:
##  randomForest(formula = kept_status ~ patient_age + remind_call_result +      provider_specialty + b
##                Type of random forest: classification
##                Number of trees: 250
## No. of variables tried at each split: 3
##
##                OOB estimate of  error rate: 22.28%
## Confusion matrix:
##                Kept Missed class.error
## Kept      22465    8594  0.2766992
## Missed    5244   25815  0.1688399

plot(rf)

```



```
varImpPlot(rf)
```



Using caret Package

```
# Look at number of cvs and repeats for faster run-time
control <- caret::trainControl(method = "cv", number = 3, allowParallel = TRUE)
```

```

seed <- 7
metric <- "Accuracy"
set.seed(seed)
mtry <- 3
tuneGrid <- expand.grid(.mtry = mtry)

#Train on subset to see how long it will take. Takes ~ 1.5 hours

rftrain <- caret::train(kept_status ~ patient_age + remind_call_result + provider_specialty + billing_t

caret::confusionMatrix(rftrain)

control <- caret::trainControl(method = "oob", number = 10, repeats = 3, search = "random")

## Warning: `repeats` has no meaning for this resampling method.

seed <- 7
metric <- "Accuracy"
set.seed(seed)
mtry <- 3
tuneGrid <- expand.grid(.mtry = mtry)

### Below code takes a long time to run, need to consider ways to shorten it
rftrain3 <- caret::train(keptstatus ~ age + remindresult + specialty +
                        billtype + hour + percent_missed + length + gender +
                        distance + new + leadtime + weekday + county,
                        data = train_balanced, method = "rf", metric = metric,
                        tuneLength = 15, trControl = control)

print(rftrain2)
plot(rf)
varImpPlot(rf)
varUsed(rf)
p_rf <- predict(rf, test)
caret::confusionMatrix(p_rf, test$keptstatus)

```