

# Exploratory Data Analysis

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## Loading the Data

Load the raw data and verify its dimensions and structure.

```
df <- readRDS('../data/tidy.Rds')
dim(df)
```

```
## [1] 1023 35
```

```
str(df)
```

```
## 'data.frame': 1023 obs. of 35 variables:
## $ acq_12_wo_or_less : logi FALSE TRUE FALSE TRUE FALSE FALSE ...
## $ age_yrs : int 7 9 10 5 5 4 6 1 8 11 ...
## $ neutered : logi TRUE TRUE TRUE TRUE TRUE TRUE ...
## $ train_6mo_or_less : logi FALSE FALSE FALSE TRUE FALSE FALSE ...
## $ train_class_count : Ord.factor w/ 4 levels "1-3"<"4-6"<"7-9"<...: NA NA NA 4 NA NA NA NA NA 3 ...
## $ train_technique : Factor w/ 2 levels "punish","reward": NA NA NA 2 NA NA NA NA NA 2 ...
## $ aggression : logi TRUE TRUE FALSE FALSE TRUE FALSE ...
## $ fear_anxiety : logi TRUE TRUE TRUE TRUE TRUE TRUE ...
## $ jumping : logi FALSE FALSE FALSE TRUE FALSE FALSE ...
## $ barking : logi FALSE FALSE FALSE FALSE TRUE FALSE ...
## $ coprophagia : logi FALSE TRUE FALSE FALSE TRUE FALSE ...
## $ compulsion : logi TRUE FALSE TRUE FALSE FALSE FALSE ...
## $ rep_materials : logi FALSE FALSE TRUE FALSE FALSE TRUE ...
## $ hyperactive : logi FALSE FALSE FALSE FALSE TRUE FALSE ...
## $ destructive : logi FALSE FALSE TRUE FALSE TRUE FALSE ...
## $ escape : logi FALSE FALSE TRUE FALSE TRUE TRUE ...
## $ mounting : logi FALSE FALSE FALSE FALSE TRUE FALSE ...
## $ owner_id : Factor w/ 669 levels "0143addbe877065bb8d940e6e8901700",...: 624 311 185 185 ...
## $ train_1_3_mo : Factor w/ 2 levels "FALSE","TRUE": NA NA NA 2 NA NA NA NA NA 1 ...
## $ train_4_mo : Factor w/ 2 levels "FALSE","TRUE": NA NA NA 1 NA NA NA NA NA 2 ...
## $ train_5_6_mo : Factor w/ 2 levels "FALSE","TRUE": NA NA NA 1 NA NA NA NA NA 1 ...
## $ train_start_age : Ord.factor w/ 3 levels "1-3 mo"<"4 mo"<...: NA NA NA 1 NA NA NA NA NA 2 ...
## $ male : logi FALSE TRUE TRUE FALSE TRUE TRUE ...
## $ device_used : logi NA NA NA TRUE NA NA ...
## $ buckle_collar : logi NA NA NA FALSE NA NA ...
## $ martingale : logi NA NA NA FALSE NA NA ...
## $ slip_collar : logi NA NA NA FALSE NA NA ...
## $ shock_collar : logi NA NA NA FALSE NA NA ...
## $ harness : logi NA NA NA TRUE NA NA ...
## $ head_halter : logi NA NA NA FALSE NA NA ...
## $ choke_collar : logi NA NA NA FALSE NA NA ...
## $ prong_collar : logi NA NA NA FALSE NA NA ...
```

```
## $ house_soiling      : logi TRUE TRUE TRUE TRUE FALSE TRUE ...
## $ adj_train_technique: Factor w/ 2 levels "punish","reward": NA NA NA 2 NA NA NA NA NA 2 ...
## $ punish_device      : Factor w/ 2 levels "FALSE","TRUE": NA NA NA 1 NA NA NA NA NA 1 ...
```

## Basic Exploration

### Owner Identifier

The owner identifier is used to calculate the number of owners.

```
# Number of unique owners after inclusion criteria.
length(unique(df$owner_id))
```

```
## [1] 641
```

It is also used to calculate the number of dogs per household.

```
summary(plyr::count(df, 'owner_id'))
```

```
##                owner_id                freq
## 0180dd62878f2d494db4e6aae4695386: 1   Min.    :1.000
## 018b0b08b0a8dbc63f58c47b0c94d2e4: 1   1st Qu.:1.000
## 01bbe34d450b00b4fc3ce4b319986b81: 1   Median  :1.000
## 01f09881fb038ee28ab0ef02aa80d87a: 1   Mean    :1.596
## 01f5bf70d07b1e05f2c5ba2fdb6c40fb: 1   3rd Qu.:2.000
## 02c983a2def62e515889b3f6657b212c: 1   Max.    :8.000
## (Other)                        :635
```

We see the median number of dogs per household is 1 (range: 1 to 8). Now we can drop the column to simplify the data set.

```
df <- subset(df, select=-c(owner_id))
```

### Overview of Data Set

Before we look at the data, let's add a basic behavior problem indicator column.

```
df <- df %>%
  mutate(behav_problem = ifelse(
    aggression | fear_anxiety | jumping | barking | coprophagia | compulsion
    | house_soiling | rep_materials | hyperactive | destructive | escape
    | mounting, TRUE, FALSE))

summary(df$behav_problem)
```

```
##      Mode   FALSE    TRUE
## logical      7    1016
```

Let's take a look at the data set as we head toward analysis.

```
summary(df)
```

```
## acq_12_wo_or_less  age_yrs      neutered      train_6mo_or_less
## Mode :logical      Min.   : 1.000    Mode :logical    Mode :logical
## FALSE:449          1st Qu.: 4.000    FALSE:132        FALSE:529
## TRUE :557           Median : 7.000    TRUE :891         TRUE :494
## NA's :17            Mean    : 7.131
##                    3rd Qu.:10.000
##                    Max.    :19.000
```

```

## train_class_count train_technique aggression      fear_anxiety
## 1-3 : 49          punish: 54      Mode :logical  Mode :logical
## 4-6 :120          reward:440      FALSE:474      FALSE:310
## 7-9 : 72          NA's :529      TRUE :549      TRUE :713
## 10+ :242
## NA's:540
##
## jumping          barking          coprophagia      compulsion
## Mode :logical    Mode :logical    Mode :logical    Mode :logical
## FALSE:793        FALSE:806        FALSE:642        FALSE:769
## TRUE :230        TRUE :217        TRUE :381        TRUE :254
##
##
##
## rep_materials    hyperactive      destructive      escape
## Mode :logical    Mode :logical    Mode :logical    Mode :logical
## FALSE:595        FALSE:907        FALSE:892        FALSE:793
## TRUE :428        TRUE :116        TRUE :131        TRUE :230
##
##
##
## mounting          train_1_3_mo      train_4_mo      train_5_6_mo      train_start_age
## Mode :logical    FALSE:248        FALSE:267        FALSE:256        1-3 mo:234
## FALSE:833        TRUE :234        TRUE :215        TRUE :226        4 mo :130
## TRUE :190        NA's :541        NA's :541        NA's :541        5-6 mo:118
##                                     NA's :541
##
##
##
## male              device_used      buckle_collar    martingale
## Mode :logical    Mode :logical    Mode :logical    Mode :logical
## FALSE:526        FALSE:62         FALSE:259        FALSE:404
## TRUE :497        TRUE :432        TRUE :235        TRUE :90
##                 NA's :529        NA's :529        NA's :529
##
##
##
## slip_collar       shock_collar      harness          head_halter
## Mode :logical    Mode :logical    Mode :logical    Mode :logical
## FALSE:449        FALSE:485        FALSE:345        FALSE:468
## TRUE :45         TRUE :9         TRUE :149        TRUE :26
## NA's :529        NA's :529        NA's :529        NA's :529
##
##
##
## choke_collar      prong_collar      house_soiling    adj_train_technique
## Mode :logical    Mode :logical    Mode :logical    punish:178
## FALSE:467        FALSE:461        FALSE:225        reward:316
## TRUE :27         TRUE :33         TRUE :798        NA's :529
## NA's :529        NA's :529
##
##
##
## punish_device     behav_problem
## FALSE:316        Mode :logical
## TRUE :178        FALSE:7
## NA's :529        TRUE :1016
##

```

```
##  
##
```

Notable observations:

- Median dog age is 7 yrs (range: 1 to 19 yrs).
  - More than half (54.4%) were acquired at 12 weeks or less.
- A majority (87.1%) of dogs were neutered.
- The gender split is nearly even with 48.6% males.
- About half of the dogs (48.3%) attended training at 6 months old or earlier (i.e., puppy training).
  - About half (47.4%) of which started attending in the 1-3 month range.
  - A majority (87.4%) of the dogs that attended puppy training were subject to some form of restraining device.
    - \* The buckle collar was the most popular device at 47.6% usage.
    - \* The shock collar was the least popular at 1.8% usage.
  - A vast majority (89.1%) were believed to have been subjected to reward-based training.
    - \* Correcting for punishing restraint devices, only 64.0% were truly subject to reward based training; a 25.1% difference!
- A vast majority of dogs (99.3%) were reported to exhibit at least one type of problematic behavior.
  - The top 3 most frequent behavior problems were house soiling, fear/anxiety, aggression.
  - The 3 least frequent behavior problems were hyperactivity, destruction, and mounting.

We create pairwise scatter plots for columns that all participants were presented (i.e., no NA responses) and we exclude the individual behavior problem columns for brevity.

```
df %>%  
  ggpairs(columns=c('acq_12_wo_or_less', 'age_yrs', 'neutered',  
                    'train_6mo_or_less', 'male'),  
    mapping=ggplot2::aes(color=behav_problem),  
    diag=list(discrete='barDiag',  
              continuous=wrap('densityDiag', alpha=0.5)),  
    legend=1,  
    progress=FALSE) +  
  theme(legend.position='bottom')
```

```
## Warning: Removed 17 rows containing non-finite values (stat_g_gally_count).
```

```
## Warning: Removed 17 rows containing non-finite values (stat_g_gally_count).
```

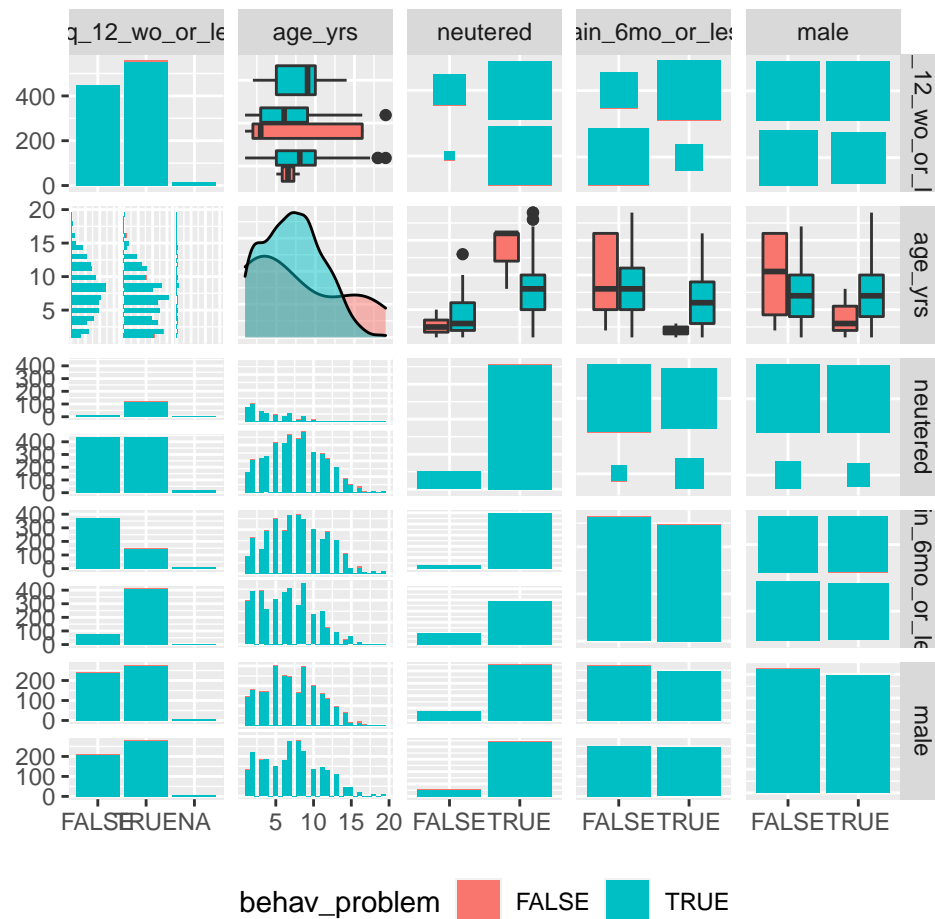
```
## Warning: Removed 17 rows containing non-finite values (stat_g_gally_count).
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
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```

```
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```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

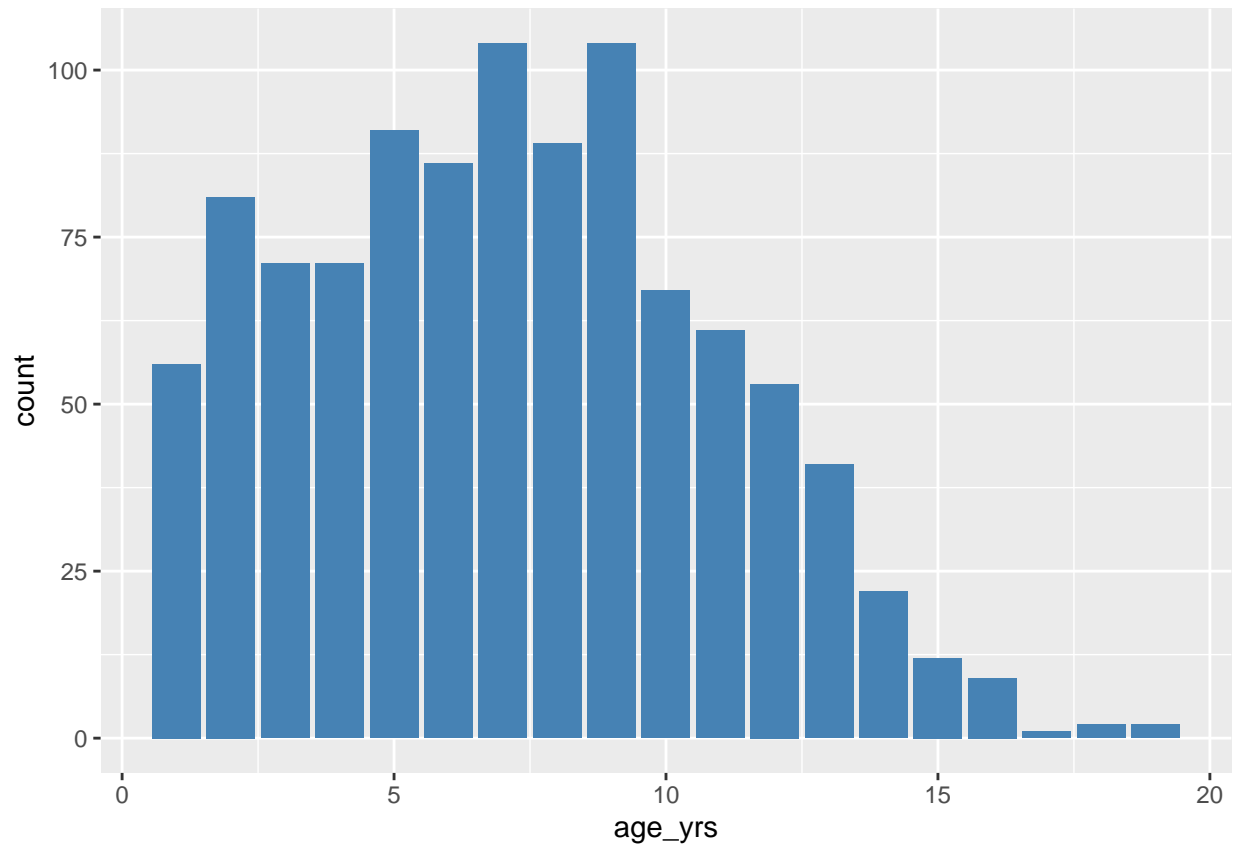


Our control group consists of the dogs that did not attend puppy training. We can compare variable distributions across the experimental and control groups by looking at the graphs along the `train_6mo_or_less` row. Thankfully, we see that the distributions between the two groups for the plotted columns are roughly equivalent.

## Continuous Variables

The age of the dog is the only continuous variable we are working with.

```
ggplot(df, aes(age_yrs)) + geom_bar(fill='steelblue')
```

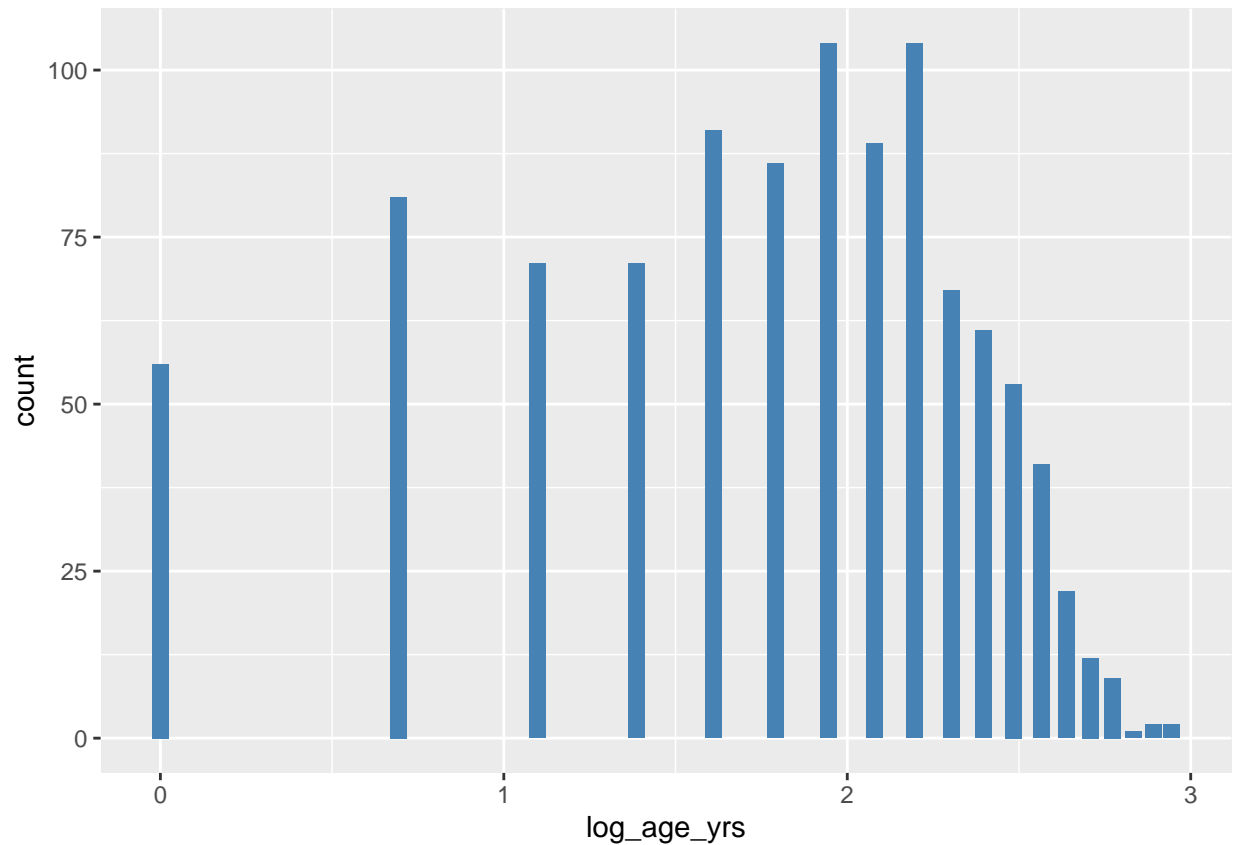


```
skewness(df$age_yrs)
```

```
## [1] 0.2669543
```

We see a slight right skew in the plot. Let's try to center it by applying a log transform.

```
df <- df %>%  
  mutate(log_age_yrs = log(age_yrs))  
ggplot(df, aes(log_age_yrs)) + geom_bar(fill='steelblue')
```



```
skewness(df$log_age_yrs)
```

```
## [1] -0.9868858
```

```
df <- subset(df, select=-c(log_age_yrs))
```

We see that the log transform resulted in a greater absolute skew, so we drop the transformed column and rely on the original.

## Discrete Variables

### Independent Variables

```
vars <- c(
  'acq_12_wo_or_less',
  'neutered',
  'train_6mo_or_less',
  'male',
  'train_1_3_mo',
  'train_4_mo',
  'train_5_6_mo',
  'train_start_age',
  'device_used',
  'buckle_collar',
  'martingale',
  'slip_collar',
  'shock_collar',
```

```

'harness',
'head_halter',
'choke_collar',
'prong_collar',
'house_soiling',
'adj_train_technique'
)

plot_list <- list()
for (i in 1:length(vars)) {
  col <- vars[i]
  p <- df %>%
    select(col) %>%
    drop_na(col) %>%
    ggplot(aes_string(x = col)) +
    geom_bar(fill='steelblue')
  plot_list[[i]] <- p
}

```

```

## Note: Using an external vector in selections is ambiguous.
## i Use `all_of(col)` instead of `col` to silence this message.
## i See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This message is displayed once per session.

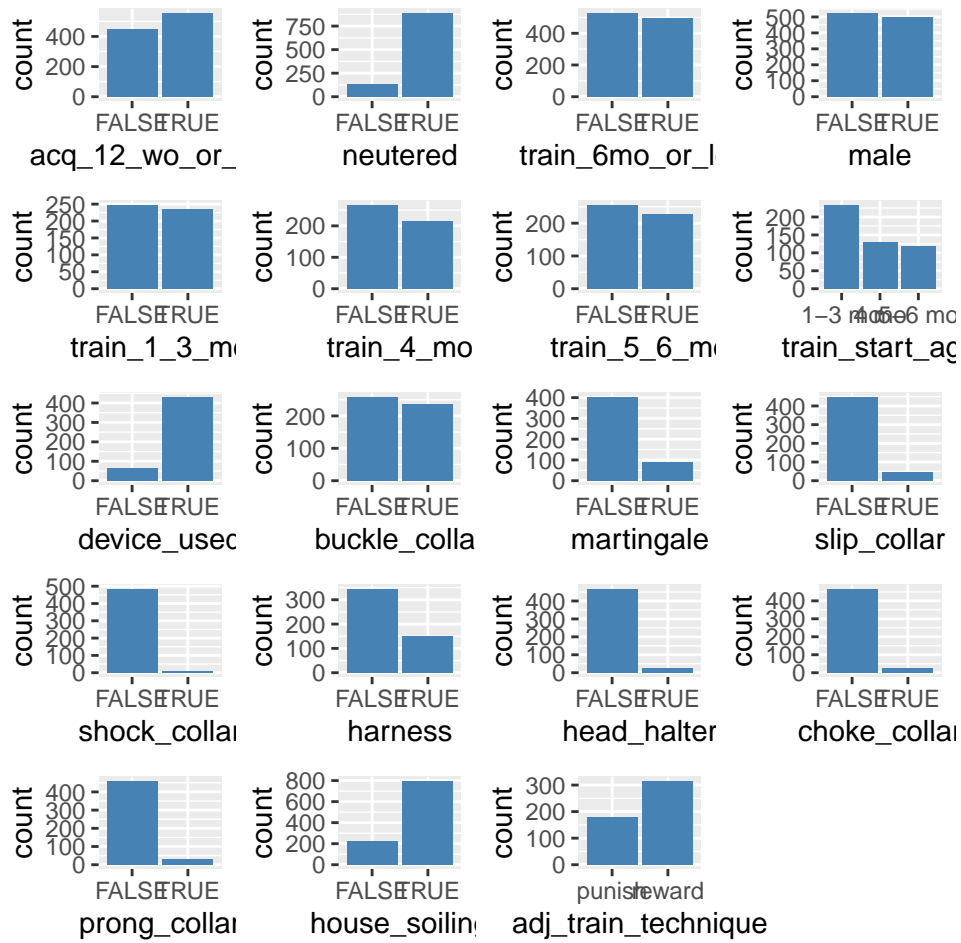
```

```

ggarrange(plotlist=plot_list, ncol=4, nrow=5)

```





## Dependent Variables

```

outcomes <- c(
  'aggression',
  'fear_anxiety',
  'jumping',
  'barking',
  'coprophagia',
  'compulsion',
  'rep_materials',
  'hyperactive',
  'destructive',
  'escape',
  'mounting',
  'house_soiling'
)

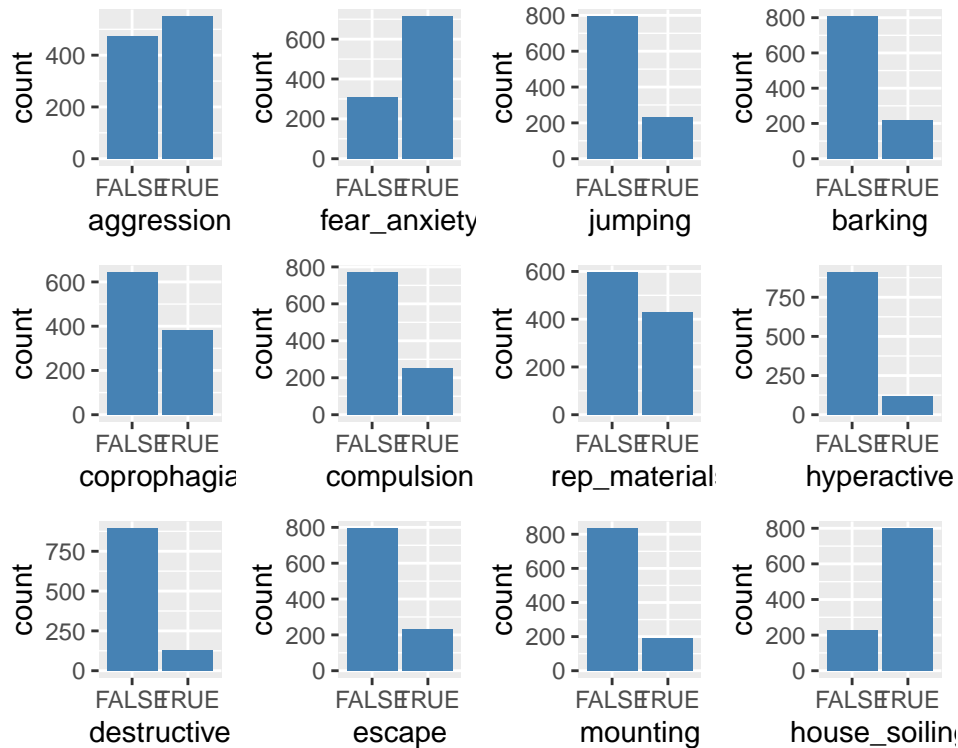
plot_list <- list()
for (i in 1:length(outcomes)) {
  col <- outcomes[i]
  p <- df %>%
    select(col) %>%

```

```

drop_na(col) %>%
  ggplot(aes_string(x = col)) +
  geom_bar(fill='steelblue')
plot_list[[i]] <- p
}
ggarrange(plotlist=plot_list, ncol=4, nrow=3)

```



## Exploring Trends and Relationships

### Sex and Neuter Status

It's common to want to know the split of neuter status by sex, so let's generate those numbers now.

```

xtab <- xtabs(~male+neutered, data=df)
print(xtab)

```

```

##           neutered
## male    FALSE TRUE
##  FALSE    76  450
##   TRUE    56  441

```

### Control vs Experimental Group

Our control group consists of the dogs who did not attend puppy training and our experimental group consists of those who did. Let's look at the variables common to both groups with the plot color indicating the presence of a behavior problem. Since we know a vast majority of dogs have at least one behavior problem, we need to look for trends in individual behavior problems for the plots to be useful.

```

# Generate plots for each attribute split by a simple predictor.
pred <- 'train_6mo_or_less'
attribs <- c(
  'acq_12_wo_or_less',
  'age_yrs',
  'male',
  'neutered'
)
attribs <- sort(attribs)
outcomes <- sort(outcomes)

plot_list <- list()
cnt <- 1
labels <- NULL
for (i in 1:length(outcomes)) {
  outcome <- outcomes[i]
  for (j in 1:length(attribs)) {
    attrib <- attribs[j]
    p <- df %>%
      drop_na(attrib) %>%
      select(attrib, outcome, pred) %>%
      ggplot(aes_string(x=attrib, fill=pred)) +
      geom_bar(position = position_dodge(0.9)) +
      labs(fill=pred) +
      theme(legend.position='none') +
      facet_grid(as.formula(paste0('~.', outcome)))
    plot_list[[cnt]] <- p
    cnt <- cnt + 1
    labels <- c(labels, outcome)
  }
}

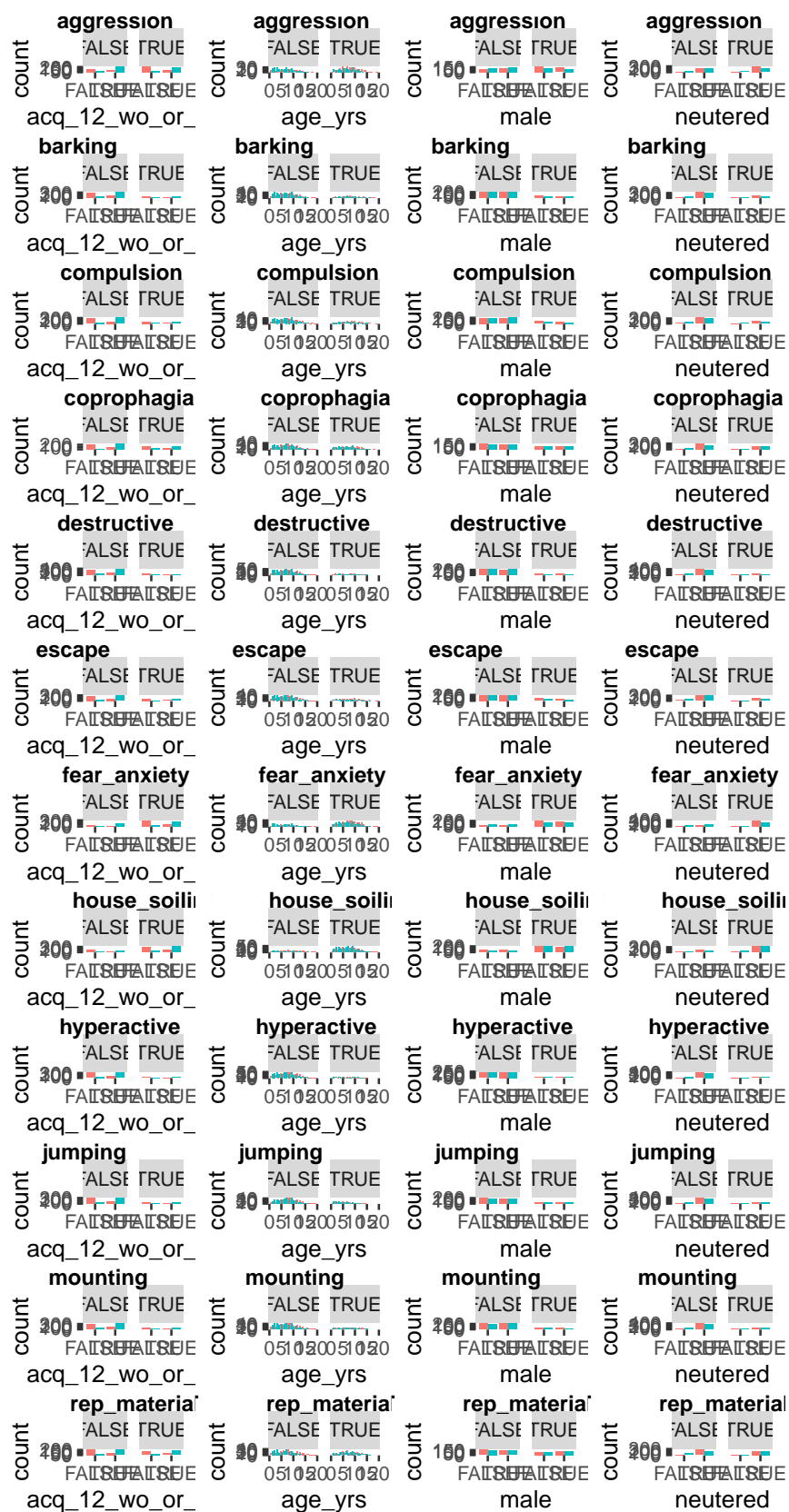
## Note: Using an external vector in selections is ambiguous.
## i Use `all_of(attrib)` instead of `attrib` to silence this message.
## i See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This message is displayed once per session.

## Note: Using an external vector in selections is ambiguous.
## i Use `all_of(outcome)` instead of `outcome` to silence this message.
## i See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This message is displayed once per session.

## Note: Using an external vector in selections is ambiguous.
## i Use `all_of(pred)` instead of `pred` to silence this message.
## i See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This message is displayed once per session.

ggarrange(plotlist=plot_list, ncol=4, nrow=12, common.legend=TRUE,
  font.label=list(size=10), vjust=0.75, legend='bottom', labels=labels)

```



train\_6mo\_or\_less FALSE TRUE

Note: For each single plot the behavior problem is indicated by the label in the top left corner. The left facet is the group of dogs without the behavior problem and the right are the dogs with the behavior problem. Within each facet the color indicates control (red) or experiment (blue) grouping.

## Within the Experimental Group

Within the experimental group we are curious to see the impact of various training techniques and restrain devices on behavior problem occurrence. We start by isolating the experimental group.

```
df_exp <- df %>%
  filter(train_6mo_or_less == TRUE)
summary(df_exp)
```

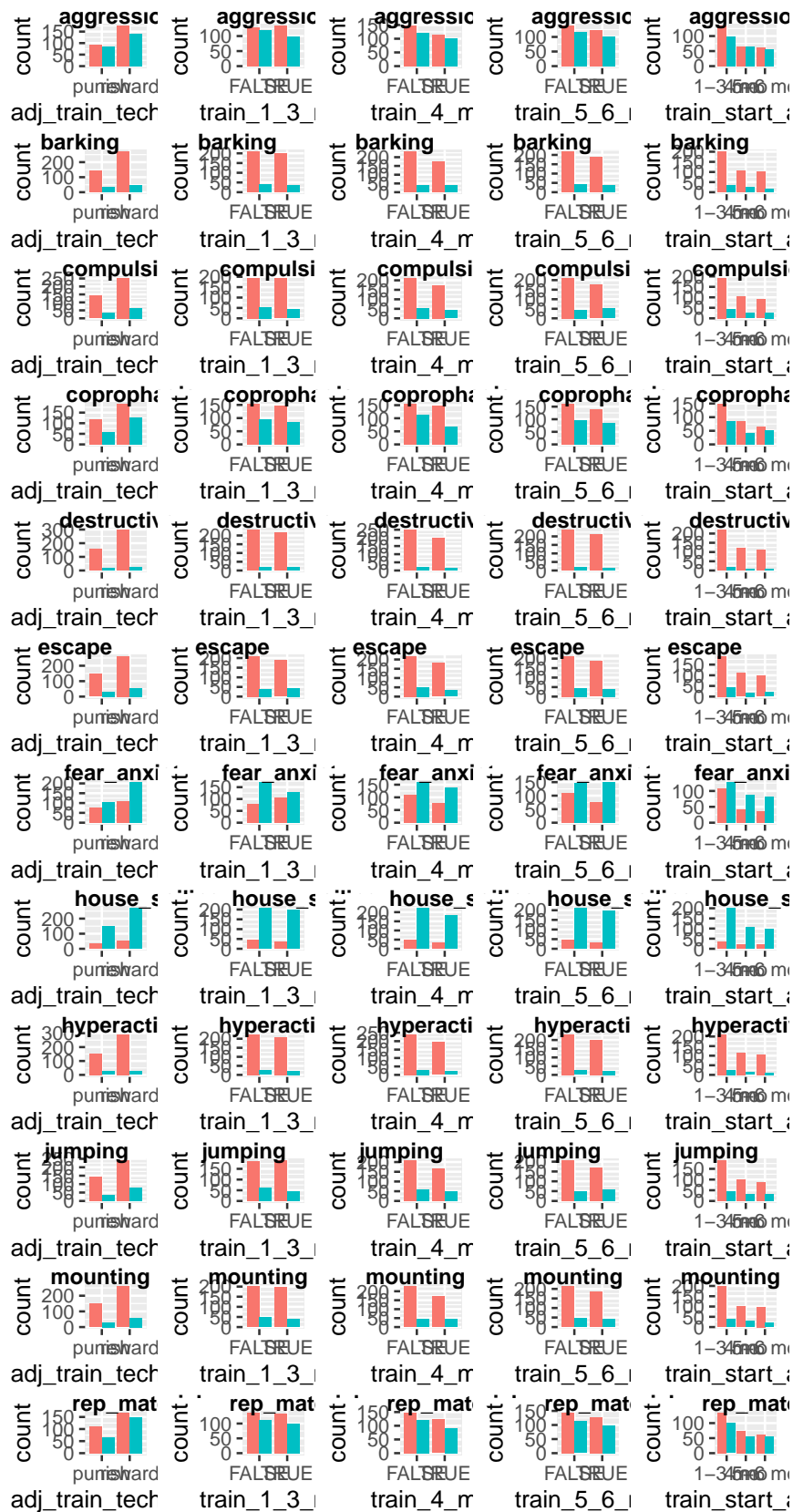
```
##  acq_12_wo_or_less    age_yrs      neutered    train_6mo_or_less
##  Mode :logical      Min.   : 1.000    Mode :logical    Mode:logical
##  FALSE:78          1st Qu.: 3.000    FALSE:103        TRUE:494
##  TRUE :410          Median : 6.000    TRUE :391
##  NA's :6            Mean   : 6.368
##                      3rd Qu.: 9.000
##                      Max.   :16.000
##  train_class_count  train_technique aggression    fear_anxiety
##  1-3 : 49           punish: 54      Mode :logical    Mode :logical
##  4-6 :120           reward:440    FALSE:267        FALSE:186
##  7-9 : 72                                TRUE :227         TRUE :308
##  10+ :242
##  NA's: 11
##
##  jumping           barking           coprophagia      compulsion
##  Mode :logical     Mode :logical    Mode :logical    Mode :logical
##  FALSE:382         FALSE:412        FALSE:308         FALSE:394
##  TRUE :112         TRUE :82         TRUE :186         TRUE :100
##
##
##  rep_materials      hyperactive      destructive      escape
##  Mode :logical      Mode :logical    Mode :logical    Mode :logical
##  FALSE:278          FALSE:442        FALSE:455        FALSE:407
##  TRUE :216          TRUE :52         TRUE :39         TRUE :87
##
##
##  mounting           train_1_3_mo    train_4_mo    train_5_6_mo    train_start_age
##  Mode :logical      FALSE:248      FALSE:267      FALSE:256      1-3 mo:234
##  FALSE:405          TRUE :234      TRUE :215      TRUE :226      4 mo :130
##  TRUE :89           NA's : 12      NA's : 12      NA's : 12      5-6 mo:118
##                                     NA's : 12
##
##
##  male           device_used      buckle_collar    martingale
##  Mode :logical   Mode :logical    Mode :logical    Mode :logical
##  FALSE:248       FALSE:62         FALSE:259        FALSE:404
##  TRUE :246       TRUE :432        TRUE :235        TRUE :90
##
##
```

```
##
## slip_collar      shock_collar      harness      head_halter
## Mode :logical    Mode :logical    Mode :logical    Mode :logical
## FALSE:449        FALSE:485        FALSE:345        FALSE:468
## TRUE :45         TRUE :9         TRUE :149        TRUE :26
##
##
##
## choke_collar     prong_collar     house_soiling    adj_train_technique
## Mode :logical    Mode :logical    Mode :logical    punish:178
## FALSE:467        FALSE:461        FALSE:81         reward:316
## TRUE :27         TRUE :33         TRUE :413
##
##
##
## punish_device    behav_problem
## FALSE:316        Mode :logical
## TRUE :178        FALSE:2
##                  TRUE :492
##
##
##
```

Now we look at the impact of training age and frequency.

```
attribs <- c(
  'train_1_3_mo',
  'train_4_mo',
  'train_5_6_mo',
  'train_start_age',
  'adj_train_technique'
)
attribs <- sort(attribs)

plot_list <- list()
cnt <- 1
labels <- NULL
for (i in 1:length(outcomes)) {
  outcome <- outcomes[i]
  for (j in 1:length(attribs)) {
    attrib <- attribs[j]
    p <- df %>%
      drop_na(attrib) %>%
      select(attrib, outcome) %>%
      ggplot(aes_string(x=attrib, fill=outcome)) +
      geom_bar(position = position_dodge(0.9)) +
      labs(fill='has this behavior problem') +
      theme(legend.position='none')
    plot_list[[cnt]] <- p
    cnt <- cnt + 1
    labels <- c(labels, outcome)
  }
}
ggarrange(plotlist=plot_list, ncol=5, nrow=12, common.legend=TRUE,
  font.label=list(size=10), vjust=0.75, legend='bottom', labels=labels)
```



has this behaviour problem ■ FALSE ■ TRUE

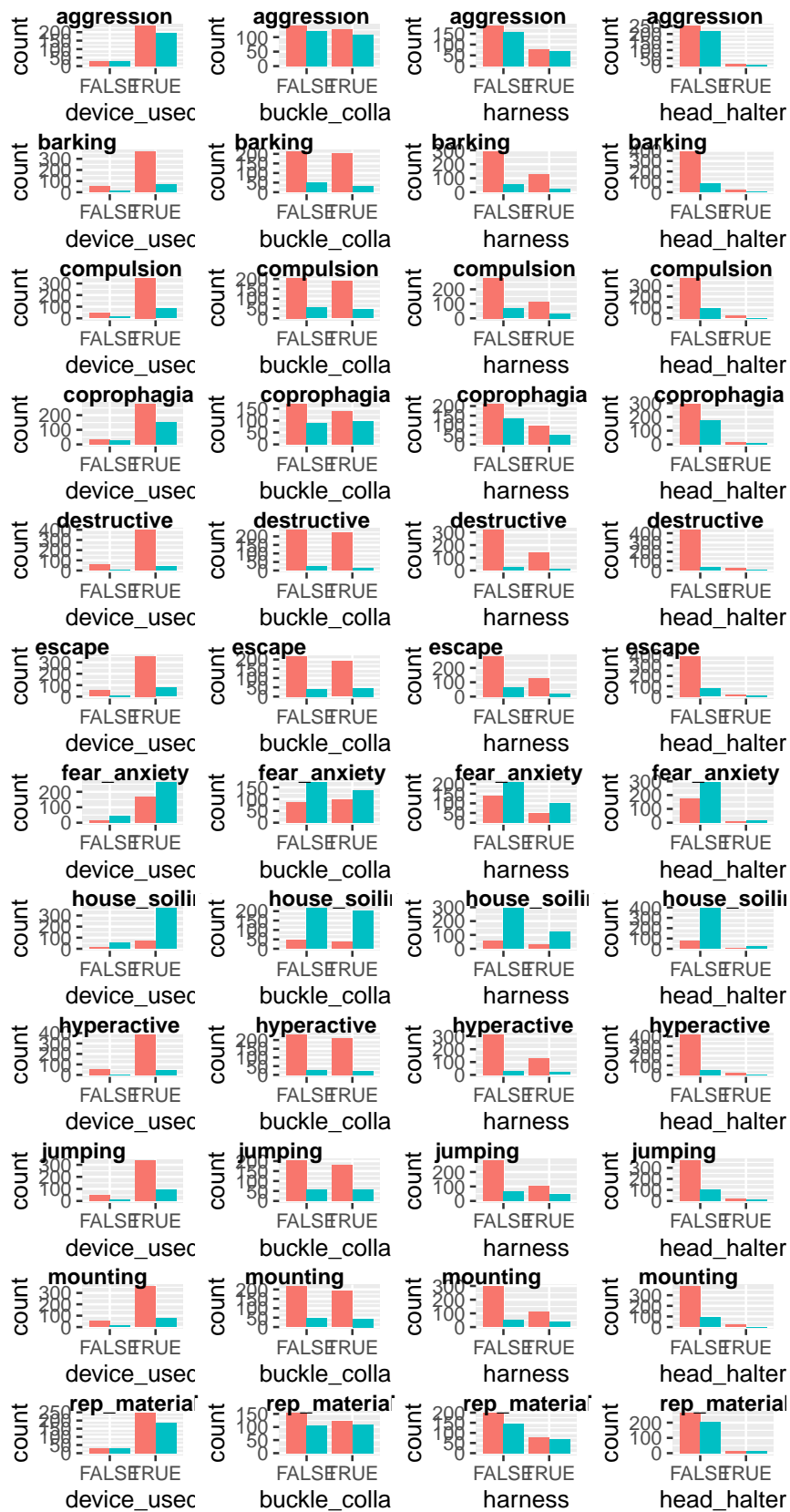
Next, we try to visual the impact

(if any) of a non-punishing restraining device choice. We also include the overall `device_used` column to see if there is a trend observed for restraining devices as a whole.

```
attribs <- c(
  'device_used',
  'buckle_collar',
  'harness',
  'head_halter'
)

plot_list <- list()
cnt <- 1
labels <- NULL
for (i in 1:length(outcomes)) {
  outcome <- outcomes[i]
  for (j in 1:length(attribs)) {
    attrib <- attribs[j]
    p <- df %>%
      drop_na(attrib) %>%
      select(attrib, outcome) %>%
      ggplot(aes_string(x=attrib, fill=outcome)) +
      geom_bar(position = position_dodge(0.9)) +
      labs(fill='has this behavior problem') +
      theme(legend.position='none')
    plot_list[[cnt]] <- p
    cnt <- cnt + 1
    labels <- c(labels, outcome)
  }
}
ggarrange(plotlist=plot_list, ncol=4, nrow=12, common.legend=TRUE,
  font.label=list(size=10), vjust=0.75, legend='bottom', labels=labels)
```



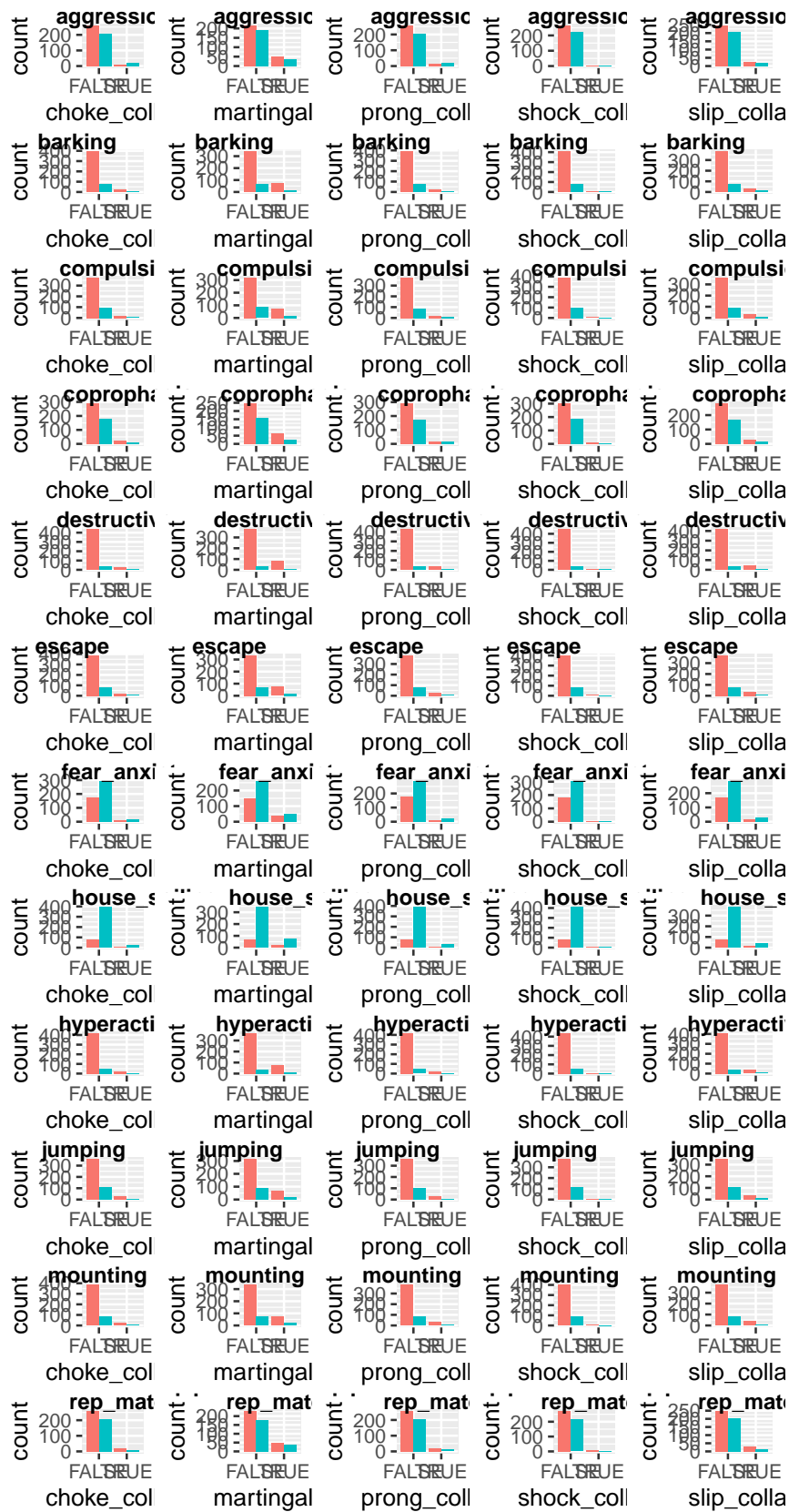


has this behaviour problem FALSE TRUE

Last, we look at the impact of punishing restraining devices.

```
attribs <- c(
  'martingale',
  'slip_collar',
  'shock_collar',
  'choke_collar',
  'prong_collar'
)
attribs <- sort(attribs)

plot_list <- list()
cnt <- 1
labels <- NULL
for (i in 1:length(outcomes)) {
  outcome <- outcomes[i]
  for (j in 1:length(attribs)) {
    attrib <- attribs[j]
    p <- df %>%
      drop_na(attrib) %>%
      select(attrib, outcome) %>%
      ggplot(aes_string(x=attrib, fill=outcome)) +
      geom_bar(position = position_dodge(0.9)) +
      labs(fill='has this behavior problem') +
      theme(legend.position='none')
    plot_list[[cnt]] <- p
    cnt <- cnt + 1
    labels <- c(labels, outcome)
  }
}
ggarrange(plotlist=plot_list, ncol=5, nrow=12, common.legend=TRUE,
  font.label=list(size=10), vjust=0.75, legend='bottom', labels=labels)
```



has this behavior problem ■ FALSE ■ TRUE