# **Exploratory Data Analysis**

This project has two raw data files at different scales from a study of infants, children, and adults watching a series of 7 video clips. I wrote Steps 1 and 2 to import and merge the data, and kept them here for your reference. Skip down to Step 3 to work on EDA.

### SOURCE DESCRIPTION

#### FILE 1: auc.csv

#### Columns:

- stim (stimulus video, levels/labels provided below)
- id (unique participant identifier)
- age (in days)
- AUC\_sal (area-under-the-curve for a saliency model)
- AUC dist (area-under-the-curve for a distance model)

AUC values indicate how well each model predicted where participants looked when watching a video. AUC values can range from 0-1 where .5 is chance and 1 is perfect prediction.

#### FILE 2: participants\_info\_full\_headers.csv

#### Columns:

- id (unique participant identifier, matches auc.csv)
- age group (a categorical age variable with levels:
- ".5-1 y" "1-1.5 y" "1.5-2 y" "2-4 y" "4-6 y" "8-10 y" "adult"
- precision (a quality measure of the eye data, smaller is better)
- 7 columns of "Seen X" the stimulus video before the study coded as SEEN (1), NOT SEEN (2), NOT SURE (3)

#### STEP 1: Read in the AUC data

Code stim as a factor.

```
auc <- read_csv(here("data_raw", "auc_bystim.csv"))
stim_levels <- 1:7
stim_labels <- c("Fallon", "Feist", "Pentatonix", "Science", "Rube", "Plane", "Dogs")
auc <- auc %>% mutate(stim = factor(stim, levels = stim_levels, labels = stim_labels))
```

## STEP 2: Read in the participant info data

Wrangle the ppt info data so that you can merge it into the auc data #Drop any data where the AUC values are missing. In the final, merged data, make the "watched" variable is coded as a factor with levels "seen" (1), "not seen" (2), "not sure" (3). Write the cleaned file to data\_cleaned/.

Read in the ppt data and rename columns to be easier to work with.

```
ppt <- read_csv(here("data_raw","participants_info_full_headers.csv")) %>%
    rename(id = `participant ID`,
        age_group = `Age group`,
        precision = "Precision")
```

Each question about watching each video is a column, so pivot\_longer. Use separate to get just the video name into it's own column.

```
ppt_long <- ppt %>%
    pivot_longer(cols = starts_with("Seen"), names_to = "stim", values_to = "watched")

ppt_long <- ppt_long %>%
    separate(stim, into = c(NA, "stim"))
```

Code stim and watched as factors.

Join the ppt data to the AUC data (by id and by stim since each participant has observations for each stim).

```
ds <- left_join(auc, ppt_long, by = c("id", "stim"))
ds <- ds %>% drop_na(AUC_sal:AUC_dist) #Drop participants for whom we don't have data for the
```

Write the data to file.

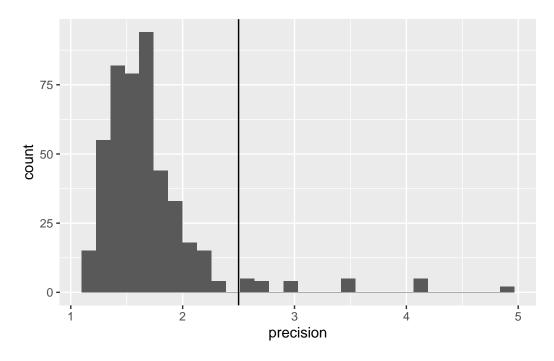
```
ds %>% write_csv(here("data_cleaned","cleaned.csv"))
```

# **STEP 3: Exploratory Data Analysis**

#### **3A Precision**

Is the precision acceptable (< 2.5) for each participant?

Visualize the distribution of precision to see if there are values above 2.5



Create a summary to figure out which participants would we need to exclude if > 2.5 meant the data are unuseable?

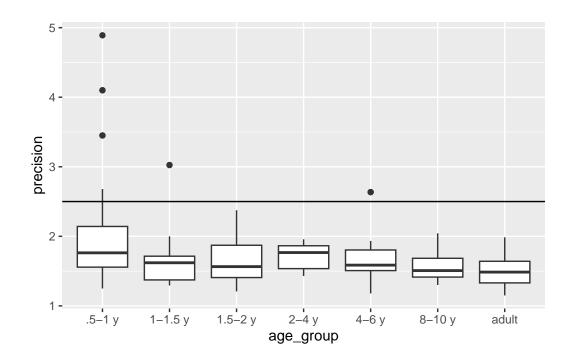
```
ds %>%
 group_by(id, age_group) %>%
 summarize(precision = mean(precision, na.RM = T)) %>%
 filter(precision > 2.5)
# A tibble: 6 x 3
# Groups: id [6]
     id age_group precision
  <dbl> <chr>
                      <dbl>
1
    52 1-1.5 y
                      3.02
    78 .5-1 y
                      3.45
2
3
    79 .5-1 y
                      4.89
4 81 .5-1 y
                      2.68
   84 4-6 у
                      2.64
5
   108 .5-1 y
                      4.1
```

Use a summary table and plots to investigate whether data equally precise for participants of different ages

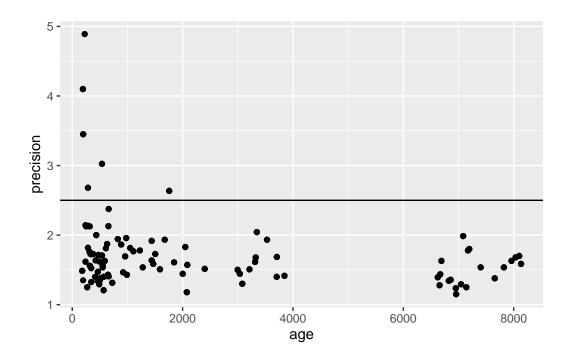
```
ds %>% group_by(age_group) %>% summarize(across(precision, list(M = mean, MIN = min, MAX = max)
```

```
# A tibble: 7 x 4
  age_group precision_M precision_MAX
  <chr>
                 <dbl>
                              <dbl>
                                            <dbl>
                                             4.89
1 .5-1 y
                  2.11
                               1.25
2 1-1.5 y
                  1.66
                               1.29
                                             3.02
                                             2.38
3 1.5-2 y
                  1.64
                               1.21
4 2-4 y
                  1.72
                               1.43
                                             1.96
                                             2.64
5 4-6 y
                  1.68
                               1.18
6 8-10 y
                  1.59
                               1.3
                                             2.04
7 adult
                  1.50
                               1.15
                                             1.99
```

```
ds %>% ggplot(aes(x = age_group, y = precision)) + geom_boxplot() + geom_hline(yintercept = 3
```



ds %>% ggplot(aes(x = age, y = precision)) + geom\_point() + geom\_hline(yintercept = 2.5)



#### 3B AGE:

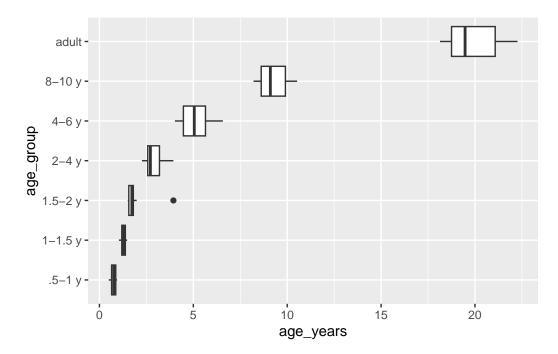
Are there any errors in age?

Convert age to years so that it can be more easily compared to age\_group

```
ds <- ds %>% mutate(age_years = age/365.25)
```

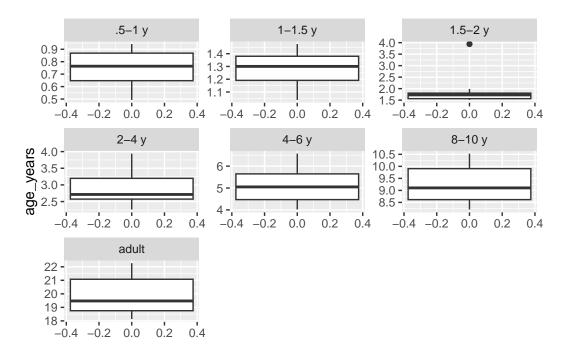
Visualize age in years by age\_group to see whether participants are the correct age for their group

```
ds %>% group_by(id, age_group) %>%
  summarize(age_years = mean(age_years)) %>%
  ggplot(aes(y = age_group, x = age_years)) + geom_boxplot()
```



Another option would be to facet by age group and to let the scales be "free" to get a better look

```
ds %>% group_by(id, age_group) %>%
  summarize(age_years = mean(age_years)) %>%
  ggplot(aes(y = age_years)) +
  geom_boxplot() +
  facet_wrap("age_group", scales = "free")
```



Make a summary table of age in years by age group to check whether all participants' ages are correct

ds %>% group\_by(age\_group) %>% summarize(min\_age = min(age\_years), max\_age = max(age\_years))

```
# A tibble: 7 x 3
  age_group min_age max_age
  <chr>
               <dbl>
                        <dbl>
1 .5-1 y
               0.493
                        0.942
2 1-1.5 y
               1.04
                        1.48
3 1.5-2 y
               1.51
                        3.94
4 2-4 y
               2.26
                        3.94
5 4-6 y
               4.02
                        6.57
6 8-10 y
               8.21
                      10.5
7 adult
                      22.3
              18.1
```