# Pilot data analysis for "Essentially blocked - Can structural factors block an essentialist interpretation of a formal explanation?"

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# Introduction

This project's repository and preregistration can be found online.

## Methods

# Planned Sample

Planned sample size is 5-year-olds and 6-year-olds recruited from a university preschool and museums

# Materials

# Procedure

2x2 design \* Context - nonstructural context, or structural context \* Explanation - control explanation, or formal explanation

Essentialism DVs \* Pilot 1 (n=14) - 5 canonical essentialism DVs (like in Muradoglu et al, other studies): stability\_past, stability\_future, innateness\_stop, inductive\_potential, innateness\_switch. But there's a worry that these items are not specific to essentialism, and a structural conception of gender may at times yield what appears to be an essentialist answer to these DVs. \* Pilot 2.1 (n=4, all structural) - testing new set of 4 DVs: fc\_expl\_group, inductivePoten, normative\_indiv, innateness\_switch \* Pilot 2.2 (n=9) - changes in group vs individual level, fc\_expl in order to make it flow better as the first item from the previous narration, normative because we're interested in group level normativity: fc\_expl\_indiv, inductivePoten, normative\_group, innateness\_switch \* Pilot 2.3 (n=7) - minor changes, fc\_expl\_indiv now mentions "size of the buckets", inductivePoten is now 4 point response instead of 2 point response: fc\_expl\_indiv\_size, inductivePoten, normative\_group, innateness\_switch

3/31 email: Nadya thinks normativity shouldn't be part of the essentialism battery, she suggests we analyze separately as exploratory. I kind of agree with her actually - discuss w Ellen

# Analysis Plan

# Exclusion criteria

Experimentor error, out of age range.

# Analysis of interest

ANOVA on essentialism measure with an interaction term for context and explanation.

We predict an interaction between context and explanation, such that in the nonstructural condition, hearing a formal explanation relative to a control explanation increases essentialism, but in the structural condition, hearing a formal explanation relative to a control explanation does not increase or increases essentialism to a lesser extent than in the nonstructural condition. In other words, the formal explanation may be interpreted structurally, and an essentialist reading blocked in the structural condition.

->

### Results

### Data preparation

Data preparation as specified in the analysis plan.

1.3.1

## v readr

```
require("knitr")
## Loading required package: knitr
## Warning: package 'knitr' was built under R version 3.5.3
#### Load relevant libraries and functions
library(tidyverse)
## -- Attaching packages -----
                               ----- tidyverse 1.2.1 --
## v ggplot2 3.1.0
                                  0.3.2
                        v purrr
## v tibble 2.0.1
                        v dplyr
                                  0.8.0.1
## v tidyr
            0.8.2
                        v stringr 1.4.0
```

v forcats 0.4.0

```
## Warning: package 'purrr' was built under R version 3.5.3
## -- Conflicts ------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(broom)
library(paramtest) # power analysis
## Warning: package 'paramtest' was built under R version 3.5.3
#### Import data for analysis
df.data <- read_csv("../data/FYP_data_pilot_children_2.csv")</pre>
## Parsed with column specification:
## cols(
##
     .default = col_character(),
     subject = col_double(),
##
##
    stim = col_double(),
##
   age = col_double(),
    fc_expl_indiv_size_expl = col_logical()
## )
## See spec(...) for full column specifications.
#### Data exclusion / filtering
# Record exclusions: experimentor error, age
df.excl <- tibble(</pre>
 exp_error = sum(df.data$exp_error == "yes"),
  age = sum(df.data$age < 5 | df.data$age >= 7)
# Exclude subjects: experimentor error, age
df.data <- df.data %>%
 filter(exp_error == "no" &
           (age >= 5 \& age < 7)) \%>\%
 select(-exp_error, -starts_with("check")) # Delete exp_error, check columns
#### Memory check analysis
#### Demographics analysis
# Add age categorical variable
df.data <- df.data %>%
  mutate(age_cat = case_when(
   age < 5 ~ "4",
   age >= 5 \& age < 6 ~ "5",
   age >= 6 \& age < 7 ~ "6",
   age >= 7 ~ "7"))
# Age
df.dem_age <- df.data %>%
```

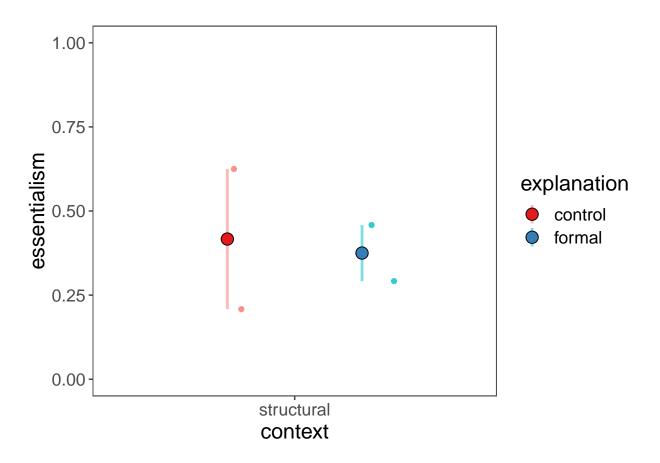
```
count(age_cat)
# Gender
df.dem_gender <- df.data %>%
  count(gender)
#### Prepare data for analysis
# Recode essentialism dependent measures
df.data$fc_expl_group <- df.data$fc_expl_group %>%
  recode("Because of the buckets in their classrooms" = 0,
         "Because girls like playing Yellow-Ball" = 1,
         "Because girls like playing Green-Ball" = 1)
df.data$fc_expl_indiv <- df.data$fc_expl_indiv %>%
  recode("Because of the buckets in her classroom" = 0,
         "Because Suzy likes playing Yellow-Ball" = 1,
         "Because Suzy likes playing Green-Ball" = 1)
df.data$fc_expl_indiv_size <- df.data$fc_expl_indiv_size %>%
  recode("Because of the size of the buckets in her classroom" = 0,
         "Because Suzy likes playing Yellow-Ball" = 1,
         "Because Suzy likes playing Green-Ball" = 1)
df.data$normative_group <- df.data$normative_group %>%
  recode("Okay" = 0,
         "Not okay" = 1)
df.data$normative_indiv <- df.data$normative_indiv %>%
  recode("Okay" = 0,
         "Not okay" = 1)
df.data$inductivePoten <-</pre>
  if_else(df.data$cb == "girlsYellow",
          (recode(df.data$inductivePoten,
                  "Green-Ball" = 1/6,
                  "Yellow-Ball" = 5/6, # conservatively estimate as midway between maybe and for sure
                  "For sure Green-Ball" = 0,
                  "Maybe Green-Ball" = 1/3,
                  "Maybe Yellow-Ball" = 2/3,
                  "For sure Yellow-Ball" = 1)),
          (recode(df.data$inductivePoten,
                  "Yellow-Ball" = 1/6,
                  "Green-Ball" = 5/6,
                  "For sure Yellow-Ball" = 0,
                  "Maybe Yellow-Ball" = 1/3,
                  "Maybe Green-Ball" = 2/3,
                  "For sure Green-Ball" = 1)))
df.data$innateness_switch <-</pre>
  if_else(df.data$cb == "girlsYellow",
          (recode(df.data$innateness_switch,
                  "For sure Green-Ball" = 0,
                  "Maybe Green-Ball" = 1/3,
                  "Maybe Yellow-Ball" = 2/3,
                  "For sure Yellow-Ball" = 1)),
```

```
(recode(df.data$innateness_switch,
                                     "For sure Green-Ball" = 1,
                                    "Maybe Green-Ball" = 2/3,
                                     "Maybe Yellow-Ball" = 1/3,
                                    "For sure Yellow-Ball" = 0)))
# Gather to tidy long form
df.tidy <- df.data %>%
    select(-starts_with("check"), -ends_with("expl")) %>%
    gather(question, response, "fc_expl_indiv":"innateness_switch")
# Order questions
df.tidy$question <- df.tidy$question %>%
    factor(levels=c('fc_expl_indiv_size', 'fc_expl_indiv', 'fc_expl_group', 'normative_indiv', 'normative
# Split dataset by DV set
df.tidy.1 <- df.tidy %>%
    filter(dvs == "fc_expl_group, inductivePoten, normative_indiv, innateness_switch")
df.tidy.2 <- df.tidy %>%
   filter(dvs == "fc_expl_indiv, inductivePoten, normative_group, innateness_switch")
df.tidy.3 <- df.tidy %>%
    filter(dvs == "fc_expl_indiv_size, inductivePoten, normative_group, innateness_switch")
df.tidy.2.3 <- df.tidy %>%
    filter(dvs == "fc_expl_indiv, inductivePoten, normative_group, innateness_switch" | dvs == "fc_expl_indiv, inductivePoten, normative_group, innateness_switch, inductivePoteness_switch, 
# Calculate overall essentialism measure per subject
df.means_subj <- df.tidy %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
df.means_subj.1 <- df.tidy.1 %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
df.means_subj.2 <- df.tidy.2 %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
df.means_subj.3 <- df.tidy.3 %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
df.means_subj.2.3 <- df.tidy.2.3 %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
df.means_subj.2.3.noNormative <- df.tidy.2.3 %>%
    filter(!str_detect(question, "normative")) %>%
    group_by(subject, age, context, explanation, dvs) %>%
    summarize(essentialism = mean(response, na.rm = TRUE))
```

# Exploring the pilot 2 data

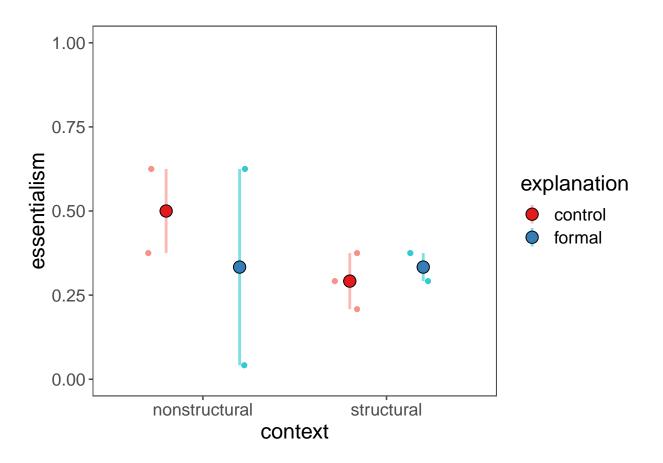
```
# Overall essentialism measure across subjects by condition
df.means <- df.tidy %>%
  group_by(context, explanation) %>%
  summarize(essentialism = mean(response, na.rm = TRUE),
           sd = sd(response, na.rm = TRUE),
           n = length(unique(subject)))
df.means
## # A tibble: 4 x 5
## # Groups: context [2]
     context
               explanation essentialism
     <chr>
                  <chr>
                                    <dbl> <dbl> <int>
## 1 nonstructural control
                                   0.354 0.389
## 2 nonstructural formal
                                   0.5 0.467
## 3 structural control
                                   0.354 0.412
## 4 structural formal
                                   0.361 0.398
                                                    6
df.means.1 <- df.tidy.1 %>%
  group_by(context, explanation) %>%
  summarize(essentialism = mean(response, na.rm = TRUE),
           sd = sd(response, na.rm = TRUE),
           n = length(unique(subject)))
df.means.1
## # A tibble: 2 x 5
## # Groups: context [1]
##
     context
             explanation essentialism
##
     <chr>>
               <chr>
                          <dbl> <dbl> <int>
## 1 structural control
                                0.417 0.454
                                                  2
## 2 structural formal
                                 0.375 0.478
df.means.2 <- df.tidy.2 %>%
  group_by(context, explanation) %>%
  summarize(essentialism = mean(response, na.rm = TRUE),
           sd = sd(response, na.rm = TRUE),
           n = length(unique(subject)))
df.means.2
## # A tibble: 4 x 5
## # Groups: context [2]
##
     context
                  explanation essentialism
     <chr>
                  <chr>
                                     <dbl> <dbl> <int>
## 1 nonstructural control
                                     0.5 0.427
## 2 nonstructural formal
                                    0.333 0.427
                                                     2
## 3 structural control
                                    0.292 0.384
                                                     3
## 4 structural
                  formal
                                    0.333 0.388
                                                     2
df.means.3 <- df.tidy.3 %>%
  group_by(context, explanation) %>%
```

```
summarize(essentialism = mean(response, na.rm = TRUE),
           sd = sd(response, na.rm = TRUE),
           n = length(unique(subject)))
df.means.3
## # A tibble: 4 x 5
## # Groups: context [2]
              explanation essentialism
##
     context
##
     <chr>
                  <chr>
                                    <dbl> <dbl> <int>
## 1 nonstructural control
                                   0.208 0.305
## 2 nonstructural formal
                                   0.667 0.471
                                    0.417 0.5
## 3 structural control
                                                     1
## 4 structural formal
                                    0.375 0.375
# Set visualization theme
theme_set(
 ggthemes::theme_few() +
 theme(text = element_text(size = 16)) # large text size for presentation figures
# Pilot 2.1: essentialism by context and explanation
ggplot(df.means_subj.1, aes(x = context, y = essentialism, fill = explanation, color = explanation)) +
 geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
            alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
              position = position_dodge(width = 0.8),
              geom = "linerange",
              size = 1,
              alpha = 0.5) +
  stat_summary(fun.y = "mean",
              position = position_dodge(width = 0.8),
              geom = "point",
              shape = 21,
              color = "black",
              size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



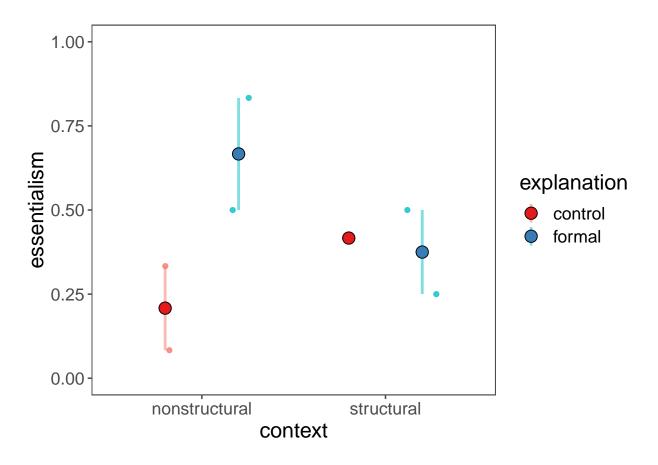
```
ggsave("FYP_pilot_children_2.1_contextExpl.png", width = 5)
```

```
# Pilot 2.2: essentialism by context and explanation
ggplot(df.means\_subj.2, aes(x = context, y = essentialism, fill = explanation, color = explanation)) +
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



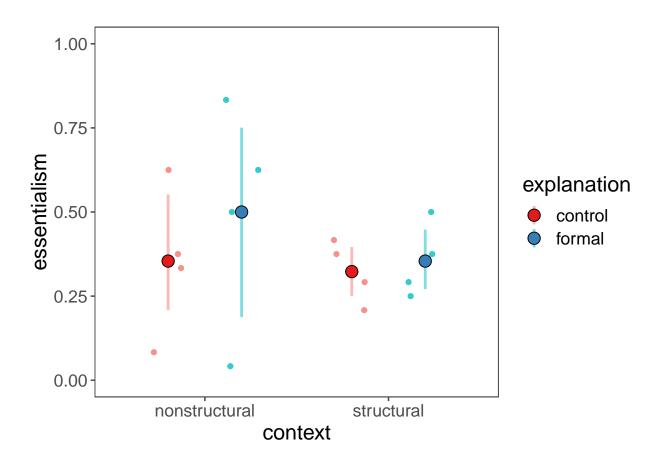
```
ggsave("FYP_pilot_children_2.2_contextExpl.png", width = 5)
```

```
# Pilot 2.3: essentialism by context and explanation
ggplot(df.means\_subj.3, aes(x = context, y = essentialism, fill = explanation, color = explanation)) +
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



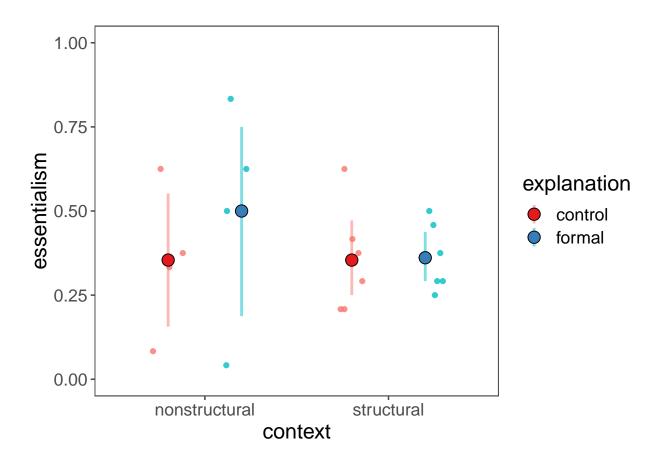
```
ggsave("FYP_pilot_children_2.3_contextExpl.png", width = 5)
```

```
\# Pilot 2.2 + 2.3: essentialism by context and explanation
ggplot(df.means_subj.2.3, aes(x = context, y = essentialism, fill = explanation, color = explanation))
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



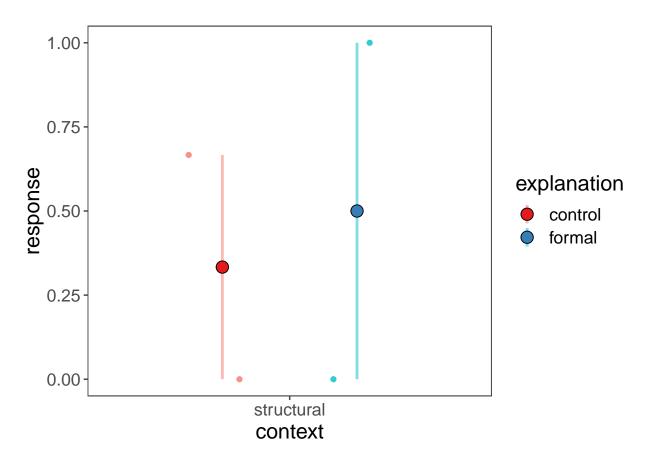
```
ggsave("FYP_pilot_children_2.2_2.3_contextExpl.png", width = 5)
```

```
# Pilot 2.1 + 2.2 + 2.3: essentialism by context and explanation
ggplot(df.means\_subj, aes(x = context, y = essentialism, fill = explanation, color = explanation)) +
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



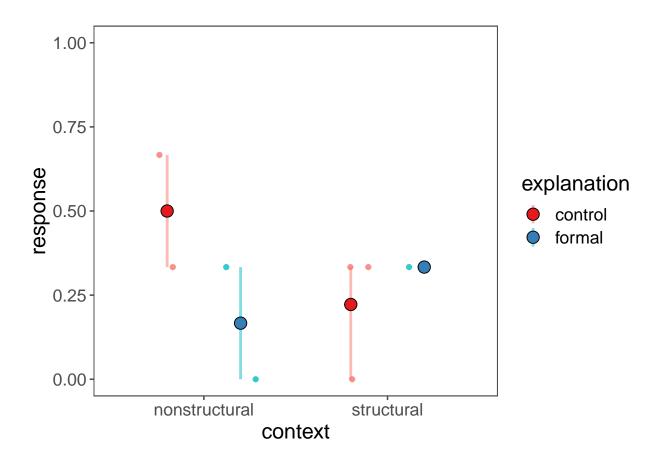
```
ggsave("FYP_pilot_children_2.1_2.2_2.3_contextExpl.png", width = 5)
```

```
# Pilot 2.1 (innateness switch only): essentialism by context and explanation
ggplot(df.tidy.1 %>% filter(question == "innateness_switch"), aes(x = context, y = response, fill = exp
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



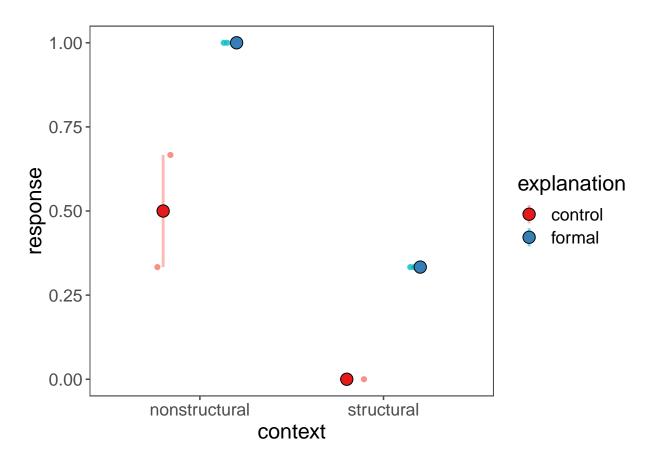
```
ggsave("FYP_pilot_children_2.1_contextExpl_innatenessSwitch.png", width = 5)
```

```
# Pilot 2.2 (innateness switch only): essentialism by context and explanation
ggplot(df.tidy.2 %>% filter(question == "innateness_switch"), aes(x = context, y = response, fill = exp
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



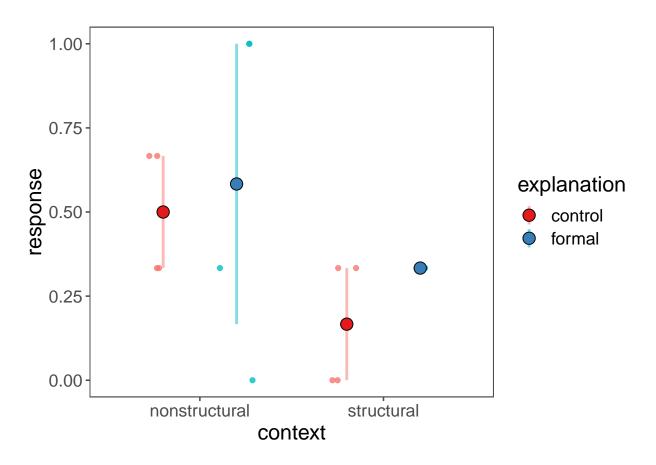
```
ggsave("FYP_pilot_children_2.2_contextExpl_innatenessSwitch.png", width = 5)
```

```
# Pilot 2.3 (innateness switch only): essentialism by context and explanation
ggplot(df.tidy.3 %>% filter(question == "innateness_switch"), aes(x = context, y = response, fill = exp
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



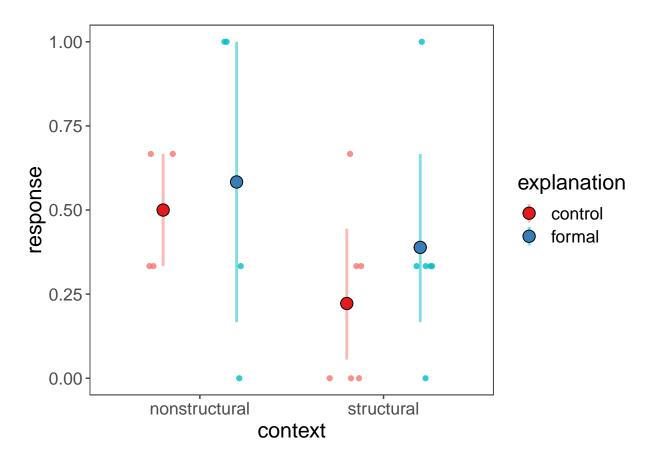
```
ggsave("FYP_pilot_children_2.3_contextExpl_innatenessSwitch.png", width = 5)
```

```
# Pilot 2.2 + 2.3 (innateness switch only): essentialism by context and explanation
ggplot(df.tidy.2.3 %>% filter(question == "innateness_switch"), aes(x = context, y = response, fill = extended filter)
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



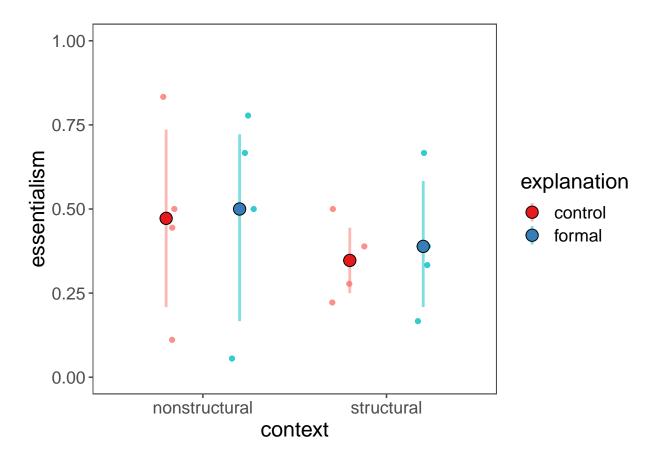
```
ggsave("FYP_pilot_children_2.2_2.3_contextExpl_innatenessSwitch.png", width = 5)
```

```
# Pilot 2.1 + 2.2 + 2.3 (innateness switch only): essentialism by context and explanation
ggplot(df.tidy %>% filter(question == "innateness_switch"), aes(x = context, y = response, fill = expla
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
              position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



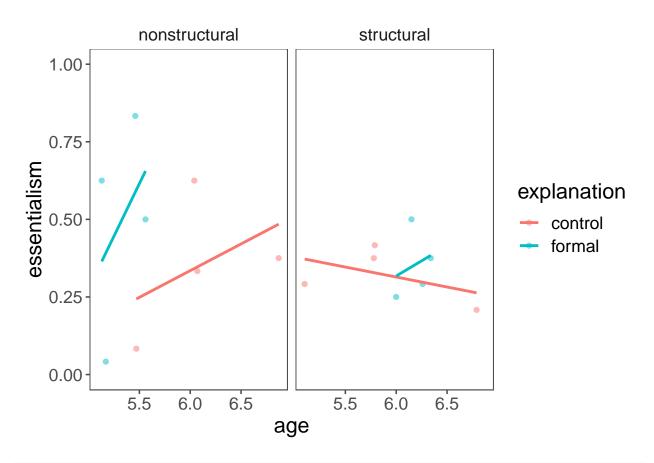
```
ggsave("FYP_pilot_children_2.1_2.2_2.3_contextExpl_innatenessSwitch.png", width = 5)
```

```
# Pilot 2.2 + 2.3 (no normative): essentialism by context and explanation
ggplot(df.means_subj.2.3.noNormative, aes(x = context, y = essentialism, fill = explanation, color = ex
  geom_point(position = position_jitterdodge(jitter.width = 0.2, dodge.width = 0.8),
             alpha = 0.8) +
  stat_summary(fun.data = "mean_cl_boot",
               position = position_dodge(width = 0.8),
               geom = "linerange",
               size = 1,
               alpha = 0.5) +
  stat_summary(fun.y = "mean",
               position = position_dodge(width = 0.8),
               geom = "point",
               shape = 21,
               color = "black",
               size = 4) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



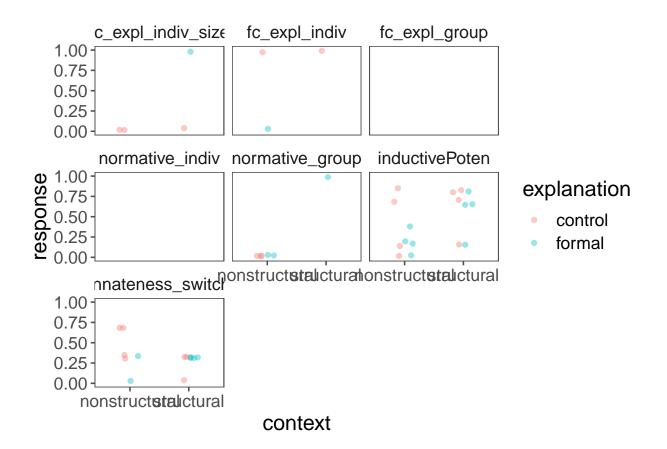
```
ggsave("FYP_pilot_children_2.2_2.3_contextExpl_noNormative.png", width = 5)
```

```
# Essentialism by explanation over age, faceted by context
ggplot(df.means_subj.2.3, aes(x = age, y = essentialism, fill = explanation, color = explanation)) +
  facet_wrap(~ context) +
  geom_point(alpha = 0.5) +
  geom_smooth(method = "lm", se = FALSE) +
  scale_fill_brewer(palette = "Set1") +
  scale_y_continuous(limits = c(0, 1))
```



```
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
```

## Warning: Removed 70 rows containing missing values (geom\_point).



# ggsave("FYP\_pilot\_children\_2\_contextExpl\_DV.png")

```
## Saving 6.5 x 4.5 in image
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in min(x): no non-missing arguments to min; returning Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning in max(x): no non-missing arguments to max; returning -Inf
## Warning: Removed 70 rows containing missing values (geom_point).
```

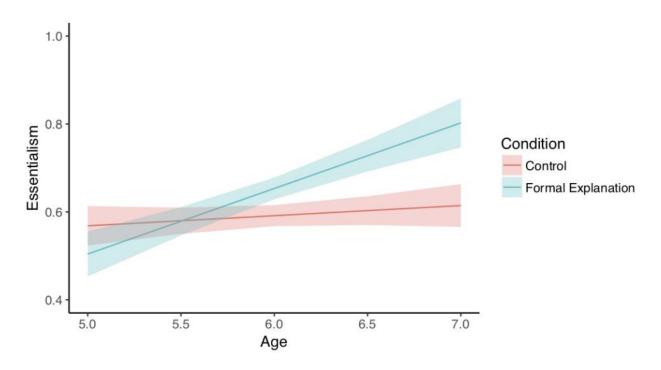


Figure 1: Muradoglu et al 2019 Exp 2 explanation on essentialism results

### Target analysis

The analyses as specified in the analysis plan.

```
# contextExpl.2.3 <- df.means_subj.2.3 %>%
# aov(essentialism ~ 1 + context * explanation, .) %>%
# tidy()
```

# Power Analysis

# Power Analysis based on prior studies

Nadya's estimate is 30 to 50 per cell, so n = 120 to 50\*4 total.

Muradoglu et al 2019 (Exp 2 on gender) ran 93 5-6yo in either a formal explanation or control explanation condition (both basically nonstructural), and measured essentialism using 5 essentialism DVs (coded as 0 to 1, with greater being more essentialist).

Although the main effect of condition was not significant, p = .11, the condition by age interaction was, p = .044. Six-year-olds, but not 5-year olds, showed higher essentialism for properties introduced via formal explanations (ps = .011 and .999, respectively; see Figure 2). Children also gave more essentialist responses with age, p = .003.

Vasilyeva et al 2018 ran 48 5-6yo (also some 4-5 and adults) in either a structural context or nonstructural context (both basically without giving an explanation), and measured essentialism using 1 essentialism DV (innateness\_switch, which they call mutability, coded as 1, 2, 3, 4, with greater being less essentialist).

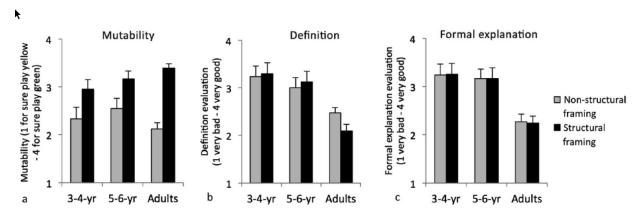


Figure 4. Mutability (a), partial definition (b), and formal explanation ratings (c) as a function of framing condition and age group. Error bars represent 1 SEM.

Figure 2: Vasilyeva et al 2019 context on mutability results

an ANOVA with framing condition and age group as between-subjects factors revealed the predicted main effect of framing, F(1, 85) = 8.95, p = .004, partial\_eta\_p\_sq = .095, with no main effect of age group, F(1, 85) = 1.05, p = .309, nor interaction, F(1, 85) = .01, p = .984. Similarly, adults rated the target property as more mutable under the structural than nonstructural framing, t(65) = 8.04, p = .001, d = 2.00.

```
# make reproducible
set.seed(1)
# set predicted values based on prior literature
predicted <- tibble(</pre>
  context = c("nonstructural", "nonstructural", "structural", "structural"),
  explanation = c("control", "formal", "control", "formal"),
  # means
  essentialism = c(0.590, \# nonstructural control \sim Muradoglu et al control condition (exact mean from
                   0.645, # nonstructural formal ~ Muradoglu et al formal condition (exact mean from da
                   1-(3.17-1)/(4-1), # structural control ~ Vasilyeva et al structural condition (eyeba
                   1-(3.17-1)/(4-1)), # structural formal ~ we predict to be the same as structural con
  # where do the SDs come into the power analysis?
  sd = c(0.475, \# nonstructural control \sim Muradoglu et al control condition (exact SD from data)
         0.463, # nonstructural formal ~ Muradoglu et al formal condition (exact SD from data)
         (0.65/3.5)/(4-1)*sqrt(24), # structural control ~ Vasilyeva et al structural condition (estima
         (0.65/3.5)/(4-1)*sqrt(24))
)
# set parameters for linear model:
\# essentialism = nonstructural + formal * x1 + structural * x2 + formalStructural * x1 * x2
b0 = predicted$essentialism[1] # b0 = nonstructural_control
b1 = predicted$essentialism[2] - b0 # nonstructural_formal = b0 + b1 -> b1 = nonstructural_formal - b0
```

b2 = predicted\$essentialism[3] - b0 # structural\_control = b0 + b2 -> b2 = structural\_control - b0

```
b3 = predicted essentialism[4] - b0 - b1 - b2 # structural_formal = <math>b0 + b1 + b2 + b3 -> b3 = structura
# make a function to simulate linear regression
regression_sim <- function(simNum, n, b0, b1, b2, b3) {</pre>
    x1 <- sample(0:1, n, replace = TRUE) # formalTRUE is either 0 or 1
    x2 <- sample(0:1, n, replace = TRUE) # structural TRUE is either 0 or 1
    # essentialism <- b0 + (b1 * formalTRUE) + (b2 * structuralTRUE) + (b3 * formalTRUE * structuralTRUE)
    y \leftarrow b0 + (b1 * x1) + (b2 * x2) + (b3 * x1 * x2) +
      # residual error
      rnorm(n, mean = 0, sd = 1) # how to set the mean and sd of residual?
    model \leftarrow lm(y \sim x1 * x2)
    summary(model)
    output <- summary(model)$coefficients</pre>
    coefs <- output[, 1]</pre>
    ps <- output[, 4]
    rsq <- summary(model)$r.squared
    results <- c(coefs, ps, rsq)
    names(results) <- c('b0_coef', 'b1_coef', 'b2_coef', 'b3_coef',</pre>
                         'b0_p', 'b1_p', 'b2_p', 'b3_p', 'rsq')
    return(results)
}
# run simulation with different ns, and different b3 (interaction effect sizes)
paramtest_results <- grid_search(regression_sim,</pre>
    params = list(n = c(100, 120, 140), # try different n
                  b3 = c(b3 * 1.1, b3, b3 * 0.9)), # try different values for b3 (context and explanati
    n.iter=1000,
    output='data.frame',
    parallel='snow',
    ncpus=3,
    b0 = b0, b1 = b1, b2 = b2)
## Running 9,000 tests...
# calculate power
power <- paramtest_results$results %>%
  group_by(n.test, b3.test) %>%
  summarize(power = sum(b3_p < .05) / n())</pre>
power
## # A tibble: 9 x 3
## # Groups: n.test [3]
   n.test b3.test power
##
      <dbl> <dbl> <dbl>
## 1
       100 -0.0605 0.055
## 2 100 -0.055 0.05
## 3 100 -0.0495 0.049
```

```
## 4
                   120 -0.0605 0.052
## 5
                   120 -0.055 0.043
## 6
                   120 -0.0495 0.05
## 7
                   140 -0.0605 0.061
## 8
                   140 -0.055 0.06
## 9
                   140 -0.0495 0.044
# # set alpha and the number of simulations
# alpha = 0.05
\# n_simulations = 1000
# # set up the simulation grid
# df.pwr = crossing(n = seq(80, 120, 1),
                                                  simulation = 1:n\_simulations,
#
                                                  alpha = alpha)
# # draw random samples from the normal distributions, save the random samples in lists like so:
# df.pwr = df.pwr \%
         group_by(n, simulation) %>% # your simulation counter
#
         mutate(nonstructural_control = rnorm(n, mean = predicted$essentialism[1], sd = predicted$sd[1]) %>%
                           nonstructural\_formal = rnorm(n, mean = predicted\$essentialism[2], sd = predicted\$sd[2]) \ \% > \% = predicted\$sd[2] + pr
#
#
                           structural\_control = rnorm(n, mean = predicted\$essentialism[3], sd = predicted\$sd[3]) \%>\% li
#
                           structural_formal = rnorm(n, mean = predicted$essentialism[4], sd = predicted$sd[4]) %>% lis
#
#
# # calculate anovas for each sample
# ## hint: group by simulation first and then nest the data before using map()
# df.pwr <- df.pwr %>%
       group_by(n, simulation) %>%
#
       nest() %>%
       mutate(anova = tidy(map(data, ~ aov(.$essentialism ~ 1 + .$context * .$explanation, .))), #this doe
#
#
                           p_value = map(anova, ~ .$p.value[3]))
#
#
#
# # calculate the proportion with which the HO would be rejected (= power)
# df.pwr %>%
       group_by(n) %>%
       summarize(power = sum(p_value < .05) / n())
```

Power is extremely low for all n and b3 tested - what's going on? Does it have to do with the mean/SD of the residual? Do I need to incorporate the SD of the predicted data somehow?

# Power Analysis based on bootstrapping pilot data

```
# pwr.t.test(d = cohensD(essentialism ~ context * explanation, data = df.means_subj),
# power = 0.8,
# sig.level = 0.05)
```

## Discussion