# **Connection to Bootstrap**

Xie and Singh also note in the aforesaid paper that a bootstrap distribution is "often an (asymptotic) confidence distribution", and a CD random variable generated from a confidence/compatibility distribution is related to and has similar theoretical properties as a bootstrap estimator.

Let's regenerate Julia's models/visualisations and compare it with our confidence distribution.

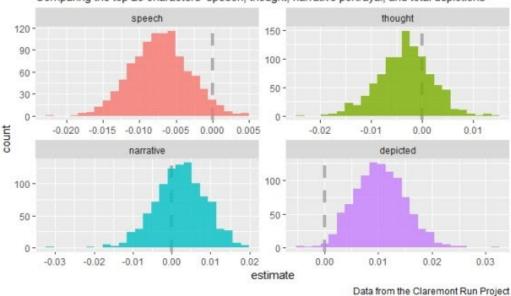
#### **Bootstrapped parameter estimates**

```
extrafont::loadfonts(device = "win")
library(tidyverse)
character visualization <- readr::read csv("https://raw.githubusercontent.com/
rfordatascience/tidytuesday/master/data/2020/2020-06-30/character visualization.csv")
xmen bechdel <- readr::read csv("https://raw.githubusercontent.com/rfordatascience/tidytuesday/
master/data/2020/2020-06-30/xmen bechdel.csv")
locations <- readr::read csv("https://raw.githubusercontent.com/rfordatascience/tidytuesday/
master/data/2020/2020-06-30/locations.csv")
per issue <- character visualization %>%
  group by(issue) %>%
  summarise(across(speech:depicted, sum)) %>%
  ungroup()
x_mansion <- locations %>%
  group by (issue) %>%
  summarise(mansion = "X-Mansion" %in% location)
locations_joined <- per_issue %>%
  inner join(x mansion)
library(tidymodels)
set.seed(123)
boots <- bootstraps(locations joined, times = 1000, apparent = TRUE)
boot models <- boots %>%
  mutate(
    model = map(
      splits,
      ~ glm(mansion ~ speech + thought + narrative + depicted,
        family = "binomial", data = analysis(.)
      )
    coef info = map(model, tidy)
  )
boot coefs <- boot models %>%
  unnest(coef info)
boot coefs %>%
  filter(term != "(Intercept)") %>%
  mutate(term = fct inorder(term)) %>%
  ggplot(aes(estimate, fill = term)) +
  geom vline (
    xintercept = 0, color = "gray50",
```

```
alpha = 0.6, lty = 2, size = 1.5
) +
geom_histogram(alpha = 0.8, bins = 25, show.legend = FALSE) +
facet_wrap(~term, scales = "free") +
labs(
   title = "Which issues contain the X-Mansion as a location?",
   subtitle = "Comparing the top 25 characters' speech, thought, narrative
portrayal, and total depictions",
   caption = "Data from the Claremont Run Project"
)
```

#### Which issues contain the X-Mansion as a location?

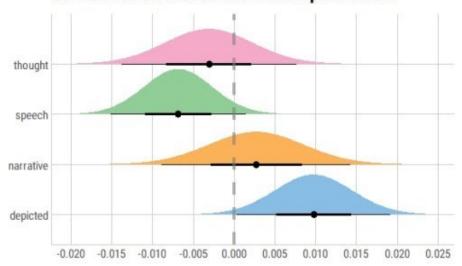




#### Confidence/Compatibility distribution based parameter estimates

```
library(ggdist)
library(hrbrthemes)
model1 <- glm(mansion ~ speech + thought + narrative + depicted,</pre>
        family = "binomial", data = locations_joined)
model1 %>%
 broom::tidy() %>%
  filter(term != "(Intercept)") %>%
  ggplot(aes(y = term, fill = term)) +
  stat dist halfeye(
    aes(dist = "student t", arg1 = df.residual(model1), arg2 = estimate, arg3 =
std.error)) +
  geom vline(xintercept = 0, color = "gray50", alpha = 0.6, lty = 2, size = 1.5)
  scale x continuous(breaks = scales::pretty breaks(n = 10)) +
  labs(x = "", y = "", fill = "Term", title = "Confidence distribution for model
parameters") +
  ggthemes::scale fill few() +
  theme ipsum rc(grid = "XY", axis = "xy") +
  theme(legend.position = "none")
```

# Confidence distribution for model parameters



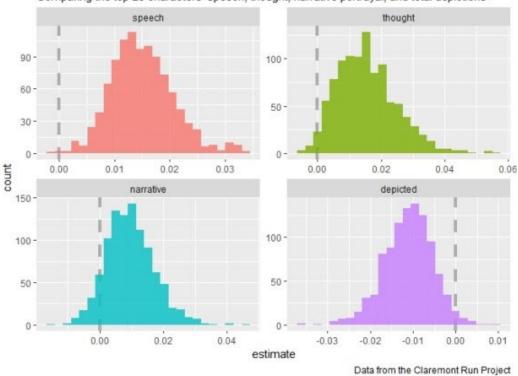
#### Very similar, if not exactly the same!

```
bechdel_joined <- per_issue %>%
  inner join(xmen bechdel) %>%
  mutate(pass bechdel = if else(pass bechdel == "yes", TRUE, FALSE))
set.seed(123)
boots <- bootstraps(bechdel joined, times = 1000, apparent = TRUE)</pre>
boot models <- boots %>%
 mutate(
    model = map(
     splits,
      ~ glm(pass_bechdel ~ speech + thought + narrative + depicted,
        family = "binomial", data = analysis(.)
      )
   ),
    coef info = map(model, tidy)
boot coefs <- boot models %>%
  unnest(coef info)
boot coefs %>%
  filter(term != "(Intercept)") %>%
  mutate(term = fct inorder(term)) %>%
  ggplot(aes(estimate, fill = term)) +
  geom vline(
    xintercept = 0, color = "gray50",
    alpha = 0.6, lty = 2, size = 1.5
  geom histogram(alpha = 0.8, bins = 25, show.legend = FALSE) +
  facet wrap(~term, scales = "free") +
```

```
labs(
   title = "Which Uncanny X-Men issues pass the Bechdel test?",
   subtitle = "Comparing the top 25 characters' speech, thought, narrative
portrayal, and total depictions",
   caption = "Data from the Claremont Run Project"
)
```

#### Which Uncanny X-Men issues pass the Bechdel test?

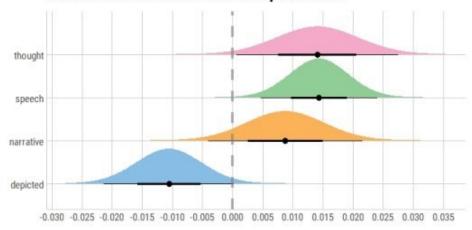




Versus...

```
model2 <- glm(pass_bechdel ~ speech + thought + narrative + depicted,</pre>
        family = "binomial", data = bechdel joined)
model2 %>%
 broom::tidy() %>%
  filter(term != "(Intercept)") %>%
  ggplot(aes(y = term, fill = term)) +
  stat dist halfeye(
    aes(dist = "student t", arg1 = df.residual(model1), arg2 = estimate, arg3 =
std.error)) +
  geom_vline(xintercept = 0, color = "gray50", alpha = 0.6, lty = 2, size = 1.5)
  scale x continuous(breaks = scales::pretty breaks(n = 10)) +
  labs(x = "", y = "", fill = "Term", title = "Confidence distribution for model
parameters") +
  ggthemes::scale fill few() +
  theme ipsum rc(grid = "XY", axis = "xy") +
  theme(legend.position = "none")
```

# Confidence distribution for model parameters

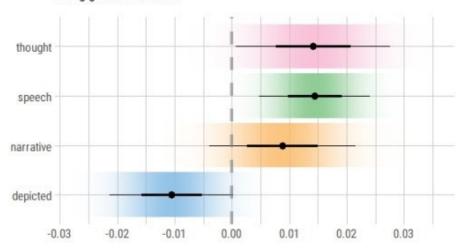


#### An alternative way of presenting this would be as follows:

```
model2 %>%
  broom::tidy() %>%
  filter(term != "(Intercept)") %>%
  ggplot(aes(y = term, fill = term)) +
  stat_dist_gradientinterval(
    aes(dist = "student_t", arg1 = df.residual(model1), arg2 = estimate, arg3 =
  std.error)) +
  geom_vline(xintercept = 0, color = "gray50", alpha = 0.6, lty = 2, size = 1.5)
+
  scale_x_continuous(breaks = scales::pretty_breaks(n = 7)) +
  labs(x = "", y = "", fill = "Term", title = "Confidence distribution of model
  parameters",
  subtitle = "Using gradient intervals") +
  ggthemes::scale_fill_few() +
  hrbrthemes::theme_ipsum_rc() +
  theme(legend.position = "none")
```

# Confidence distribution of model parameters

# Using gradient intervals



So, an interesting concept and useful alternative! Yet, the utility of ggdist is not limited to frequentist uncertainty visualisations: it also has geoms for visualising uncertainty in Bayesian models or sampling distributions.

The concept of a confidence/compatibility distribution was an interesting find for me, as somebody who was trained in ML but now prefers (and is continuously learning and trying to utilise) statistics for majority of data science projects.