Analysis

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2025-05-04

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1	Γ	Oata Loading	

```
data <- read.csv("/Users/Patron/Desktop/project/data/data_merged.csv")</pre>
```

2 Hypothesis Testing

2.1 Hypothesis 1 (H1) Trust in BioGPT vs Gemini

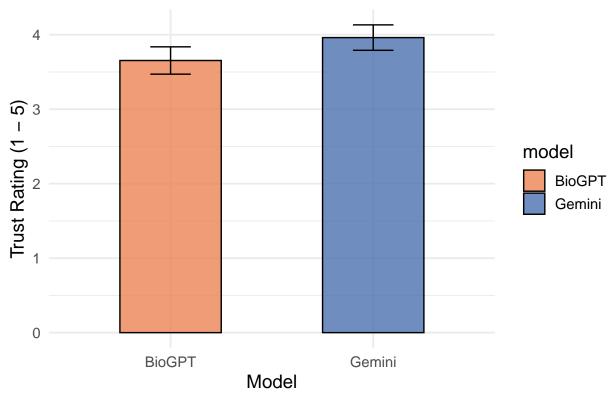
```
data <- data %>%
  mutate(
    trust_biogpt = ifelse(model_order_encoded == 0, trust_rating_B, trust_rating_A),
    trust_gemini = ifelse(model_order_encoded == 0, trust_rating_A, trust_rating_B)
)

data %>% filter(model_order_encoded == 0) %>% select(trust_rating_A, trust_rating_B)
```

```
##
      trust_rating_A trust_rating_B
## 1
                   4
## 2
                  4
## 3
                  5
                                 4
## 4
                  4
                                 5
## 5
                  5
                                 4
## 6
                  3
                                 2
## 7
                  5
## 8
                  3
                                  4
## 9
                  4
                                  3
## 10
                  3
                  5
                                 1
## 11
## 12
                  3
                                  4
                                  3
## 13
                  4
## 14
                  4
                                  3
## 15
# Normality tests
shapiro.test(data$trust_biogpt)
##
## Shapiro-Wilk normality test
## data: data$trust_biogpt
## W = 0.86394, p-value = 0.002684
shapiro.test(data$trust_gemini)
##
## Shapiro-Wilk normality test
## data: data$trust_gemini
## W = 0.85662, p-value = 0.001917
# Mann-Whitney U Test
wilcox.test(data$trust_biogpt, data$trust_gemini, paired = FALSE)
## Warning in wilcox.test.default(data$trust_biogpt, data$trust_gemini, paired =
## FALSE): cannot compute exact p-value with ties
## Wilcoxon rank sum test with continuity correction
## data: data$trust_biogpt and data$trust_gemini
## W = 280.5, p-value = 0.2684
## alternative hypothesis: true location shift is not equal to 0
# Effect size calculation
data %>%
 mutate(group = "BioGPT vs Gemini") %>%
wilcox_effsize(trust_biogpt ~ trust_gemini, paired = FALSE)
```

```
## # A tibble: 6 x 7
                  group1 group2 effsize
##
   .у.
                                           n1
                                                 n2 magnitude
                  <chr> <chr>
## * <chr>
                                  <dbl> <int> <int> <ord>
## 1 trust_biogpt 2
                         3
                                 0.624
                                            1
                                                  7 large
## 2 trust_biogpt 2
                         4
                                 0.224
                                                 10 small
## 3 trust biogpt 2
                         5
                                 0.203
                                                  8 small
                                            1
## 4 trust biogpt 3
                                 0.552
                                            7
                                                 10 large
## 5 trust_biogpt 3
                         5
                                            7
                                                  8 moderate
                                 0.406
## 6 trust_biogpt 4
                                 0.0557
                                           10
                                                  8 small
```

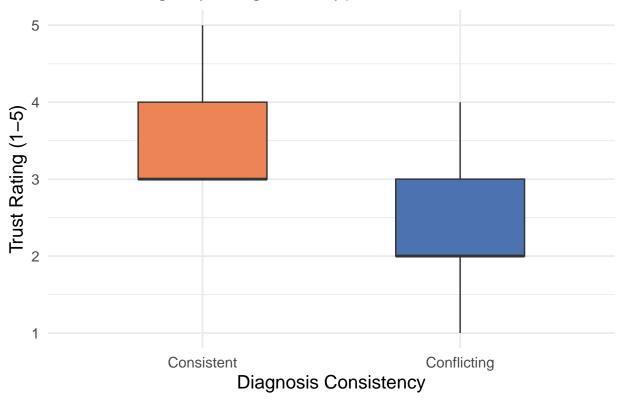
Mean Trust Ratings: BioGPT vs Gemini



2.2 Hypothesis 2 (H2) — Conflict reduces Trust

```
data$diagnosis_type_label <- factor(data$diagnosis_type_encoded,</pre>
                                    levels = c(0, 1),
                                    labels = c("Consistent", "Conflicting"))
# Normality test
shapiro.test(data$final_trust_rating)
##
##
  Shapiro-Wilk normality test
## data: data$final_trust_rating
## W = 0.89951, p-value = 0.01527
# Mann-Whitney U Test
wilcox.test(final_trust_rating ~ diagnosis_type_label, data = data)
## Warning in wilcox.test.default(x = DATA[[1L]], y = DATA[[2L]], ...): cannot
## compute exact p-value with ties
##
   Wilcoxon rank sum test with continuity correction
## data: final_trust_rating by diagnosis_type_label
## W = 139, p-value = 0.001904
## alternative hypothesis: true location shift is not equal to 0
# Effect size
data %>%
 wilcox_effsize(final_trust_rating ~ diagnosis_type_label)
## # A tibble: 1 x 7
   .у.
                        group1
                                   group2
                                               effsize
                                                          n1
                                                                n2 magnitude
                                                 <dbl> <int> <int> <ord>
## * <chr>
                        <chr>
                                   <chr>
## 1 final_trust_rating Consistent Conflicting 0.614
                                                          11
                                                                15 large
# Trust Rating Boxplot
ggplot(data, aes(x = diagnosis_type_label, y = final_trust_rating, fill = diagnosis_type_label)) +
  geom_boxplot(width = 0.5, outlier.shape = 21, outlier.fill = "white", outlier.color = "black") +
  scale_fill_manual(values = c("Consistent" = "#EC8456", "Conflicting" = "#4C72B0")) +
  labs(title = "Trust Ratings by Diagnosis Type",
       x = "Diagnosis Consistency",
      y = "Trust Rating (1-5)") +
  theme minimal(base size = 14) +
  theme(legend.position = "none")
```

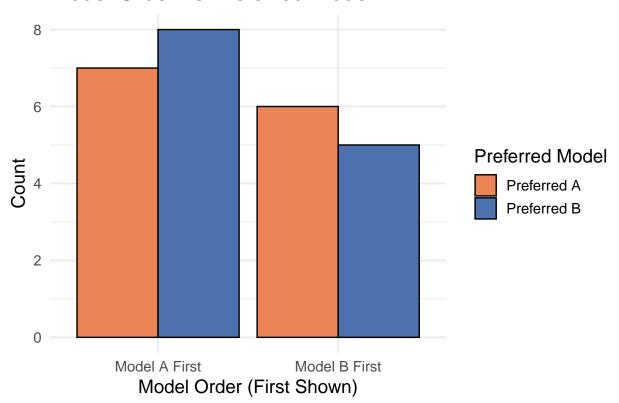
Trust Ratings by Diagnosis Type



2.3 Hypothesis 3 (H3): Anchoring Bias (Chi-square Test)

```
table(data$model_order_encoded, data$preferred_model_encoded) # View contingency table
##
##
       0 1
##
     0 7 8
##
     1 6 5
chisq.test(table(data$model_order_encoded, data$preferred_model_encoded))
##
  Pearson's Chi-squared test with Yates' continuity correction
## data: table(data$model_order_encoded, data$preferred_model_encoded)
## X-squared = 0, df = 1, p-value = 1
# Create Proper Labels
data$model_order_label <- factor(data$model_order_encoded, labels = c("Model A First", "Model B First")</pre>
data$preferred_model_label <- factor(data$preferred_model_encoded, labels = c("Preferred A", "Preferred
# Anchoring Bias Plot
```

Model Order vs Preferred Model



2.4 Hypothesis 4 (H4): Confidence — Conflict vs Consistent (Independent-samples t-test)

```
wilcox.test(final_confidence_rating ~ diagnosis_type_label, data = data)

## Warning in wilcox.test.default(x = DATA[[1L]], y = DATA[[2L]], ...): cannot

## compute exact p-value with ties

##

## Wilcoxon rank sum test with continuity correction

##

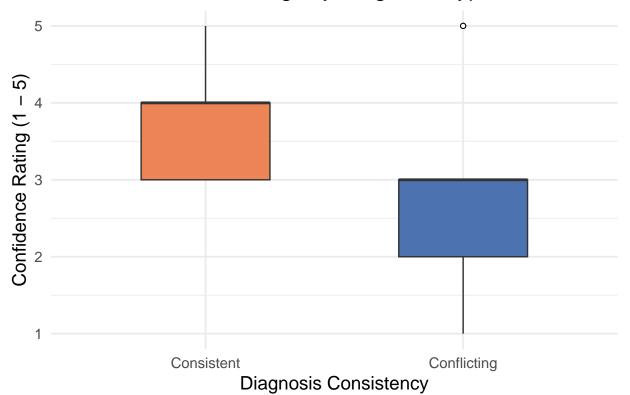
## data: final_confidence_rating by diagnosis_type_label

## W = 121, p-value = 0.03845

## alternative hypothesis: true location shift is not equal to 0
```

```
# Effect size
data %>%
 wilcox_effsize(final_confidence_rating ~ diagnosis_type_label)
## # A tibble: 1 x 7
                             group1
                                        group2
                                                                     n2 magnitude
   .у.
                                                    effsize
                                                               n1
## * <chr>
                             <chr>
                                                      <dbl> <int> <int> <ord>
                                        <chr>
## 1 final_confidence_rating Consistent Conflicting
                                                      0.411
                                                               11
                                                                     15 moderate
# Confidence Rating Boxplot
ggplot(data, aes(x = diagnosis_type_label, y = final_confidence_rating, fill = diagnosis_type_label)) +
  geom_boxplot(width = 0.5, outlier.shape = 21, outlier.fill = "white", outlier.color = "black") +
  scale_fill_manual(values = c("Consistent" = "#EC8456", "Conflicting" = "#4C72B0")) +
  labs(title = "Final Confidence Ratings by Diagnosis Type",
      x = "Diagnosis Consistency",
      y = "Confidence Rating (1 - 5)") +
  theme_minimal(base_size = 14) +
  theme(legend.position = "none")
```

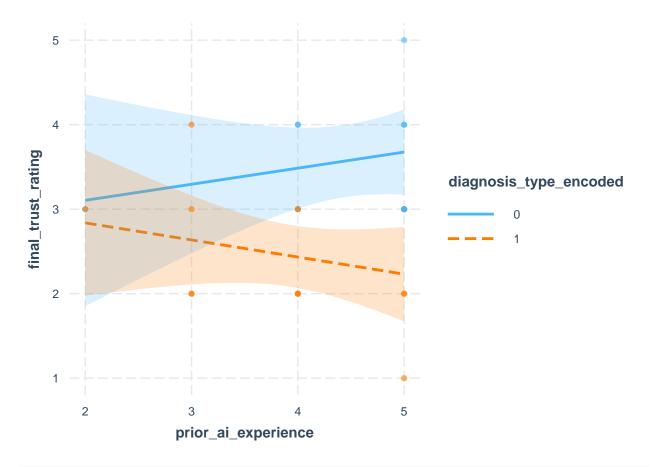
Final Confidence Ratings by Diagnosis Type



3 Linear Regression

3.1 to examine how prior experience, medical knowledge, or cognitive biases influence trust — and whether they moderate the relationship between model type or diagnosis consistency and trust.

```
data$diagnosis_type_encoded <- as.factor(data$diagnosis_type_encoded)</pre>
# Linear model with interaction
model <- lm(final_trust_rating ~ diagnosis_type_encoded * prior_ai_experience + medical_knowledge_level
           data = data)
summary(model)
##
## Call:
## lm(formula = final_trust_rating ~ diagnosis_type_encoded * prior_ai_experience +
      medical_knowledge_level, data = data)
##
## Residuals:
                               3Q
##
      Min
              1Q Median
                                      Max
## -1.0547 -0.4532 -0.1563 0.4563 1.5027
##
## Coefficients:
##
                                              Estimate Std. Error t value
## (Intercept)
                                                2.3597 1.0850 2.175
                                                           1.2972 0.400
## diagnosis_type_encoded1
                                                0.5188
## prior_ai_experience
                                                0.1903
                                                           0.2315
                                                                   0.822
## medical_knowledge_level
                                                0.1858
                                                           0.1001 1.856
## diagnosis_type_encoded1:prior_ai_experience -0.3927
                                                           0.2983 -1.316
##
                                              Pr(>|t|)
## (Intercept)
                                                0.0412 *
## diagnosis_type_encoded1
                                                0.6933
## prior_ai_experience
                                                0.4202
## medical_knowledge_level
                                                 0.0775
## diagnosis_type_encoded1:prior_ai_experience
                                                0.2022
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6802 on 21 degrees of freedom
## Multiple R-squared: 0.5105, Adjusted R-squared: 0.4172
## F-statistic: 5.475 on 4 and 21 DF, p-value: 0.003517
# Interaction plot
interact_plot(model, pred = prior_ai_experience, modx = diagnosis_type_encoded,
             plot.points = TRUE, interval = TRUE)
```



Warning: medical_knowledge_level and diagnosis_type_encoded are not included in an ## interaction with one another in the model.

