

Computational Linguistics Project

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Reading the File

```
rpp = read.csv("rpp_data.csv", header = TRUE, sep = ",")
library(dplyr)

## Warning: package 'dplyr' was built under R version 3.4.4
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
rpp = rpp %>% filter(TextNumber != "" & Abstract != "No Abstract" & Replicate..R. != "")

## Warning: package 'bindrcpp' was built under R version 3.4.4
rpp = rpp[,c(138, 139,76,24, 30, 36,37,72)]
colnames(rpp) = c("TextNumber", "Abstract", "Replicated", "Citation Count",
                  "Discipline", "SurprisingResult", "ExcitingResult",
                  "Direction of Replication")
rpp$Replicated = ifelse(rpp$Replicated == "yes", "Yes", "No")
```

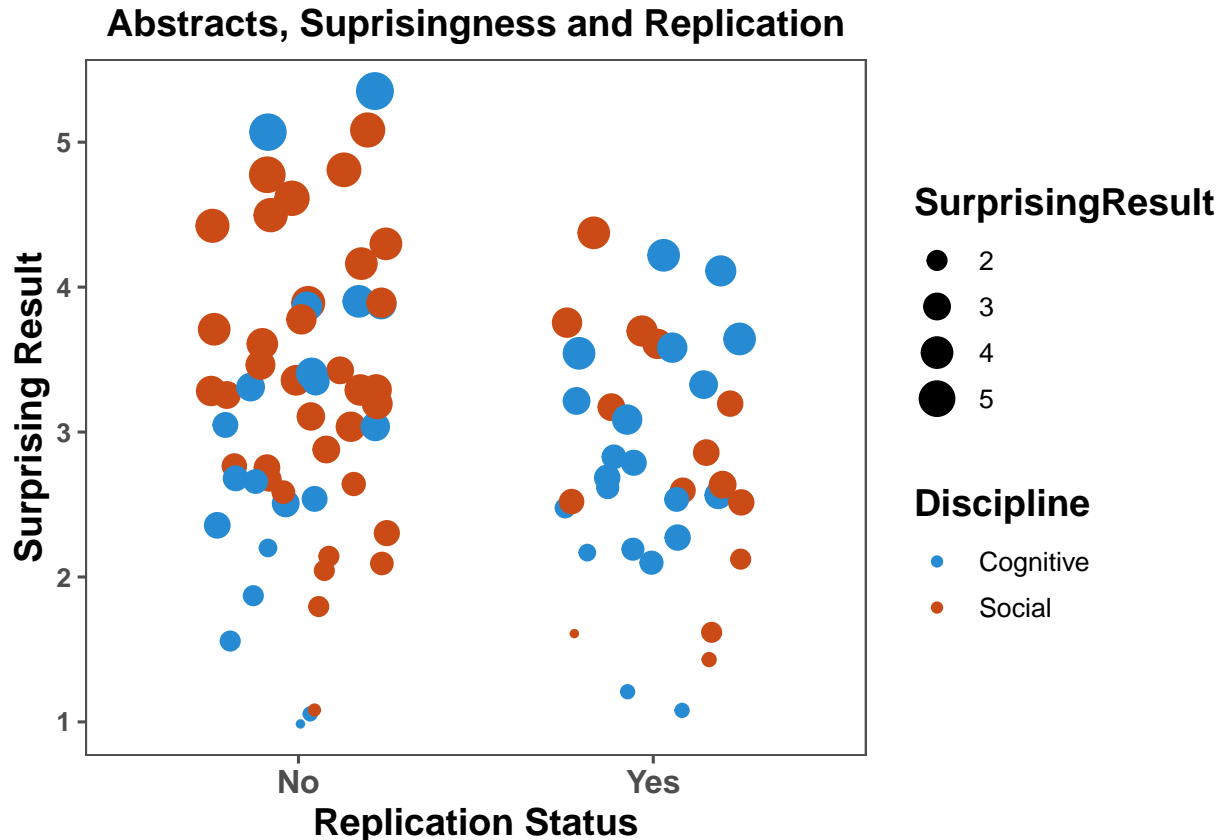
Plotting Studies

```
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 3.4.4
library(ggthemes)

## Warning: package 'ggthemes' was built under R version 3.4.4
rpp_discipline = group_by(rpp, Discipline, Replicated) %>%
  count(Studies = n())
rpp$SurprisingResult = as.numeric(as.character(rpp$SurprisingResult))
rpp %>%
  ggplot(aes(x = Replicated, y = SurprisingResult)) +
  geom_jitter(aes(color = Discipline, size = SurprisingResult), width = 0.25, height = 0.5) +
  # geom_bar(stat = "identity", position = "dodge", width = 0.5) +
  theme_few() +
  scale_color_solarized() +
  xlab("Replication Status") + ylab("Surprising Result") +
  ggtitle("Abstracts, Suprisingness and Replication") +
```

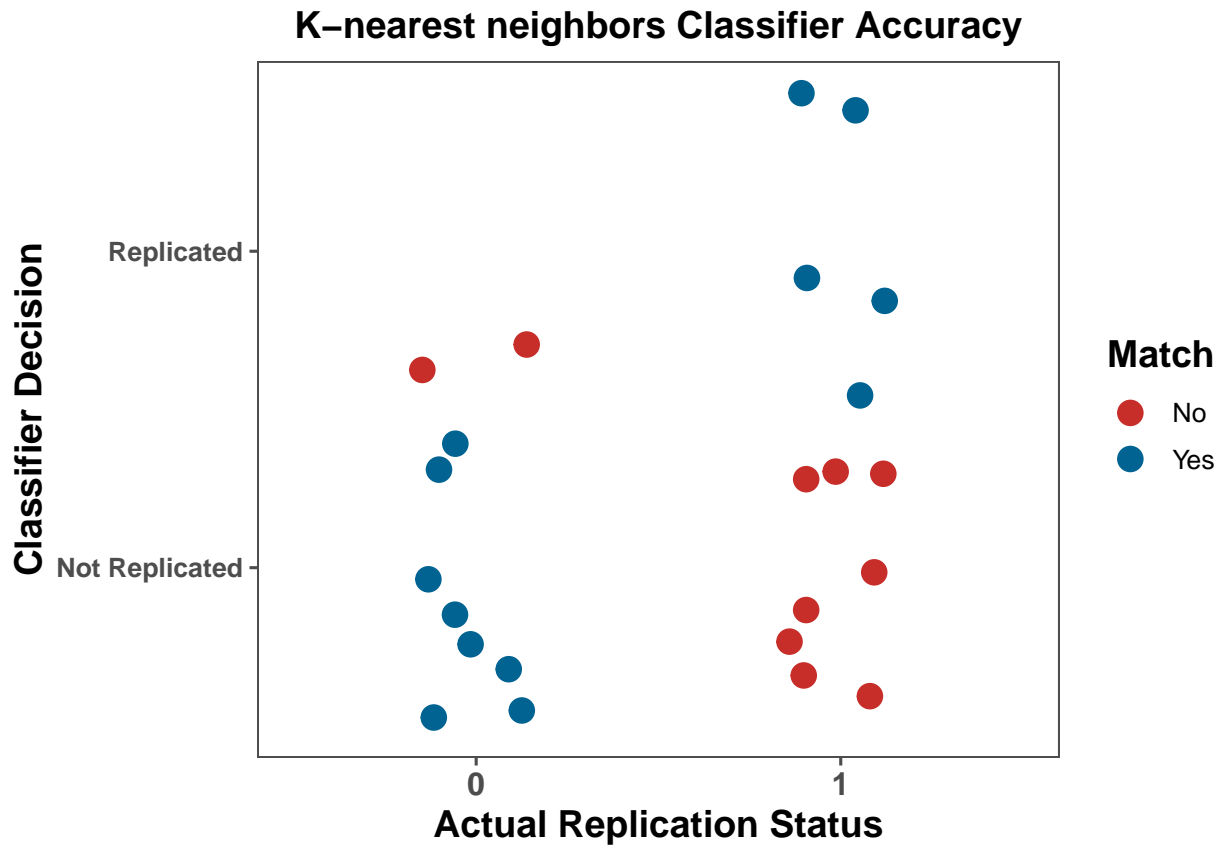
```
theme(axis.text = element_text( face = "bold", size = rel(0.8)),
      axis.title = element_text(face = "bold", size = rel(1.2)),
      legend.title = element_text(face = "bold", size = rel(1.2)),
      axis.text.x = element_text(face = "bold", size = rel(1.2)),
      plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))
```



Classifier Decisions

Nearest Neighbors

```
c3 = read.csv("classify_knn.csv", header = TRUE, sep = ",")
c3 %>%
  ggplot(aes(x =factor(Actual), y = Decision)) +
    geom_jitter( width = 0.15, height = 0.5, aes(color = Match), size = 4)+
    theme_few()+
    scale_color_wsj()+
    xlab("Actual Replication Status") + ylab("Classifier Decision") +
    ggtitle("K-nearest neighbors Classifier Accuracy") +
    theme(axis.text = element_text( face = "bold", size = rel(0.8)),
          axis.title = element_text(face = "bold", size = rel(1.2)),
          legend.title = element_text(face = "bold", size = rel(1.2)),
          axis.text.x = element_text(face = "bold", size = rel(1.2)),
          plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))
```



Classifier Performance

```

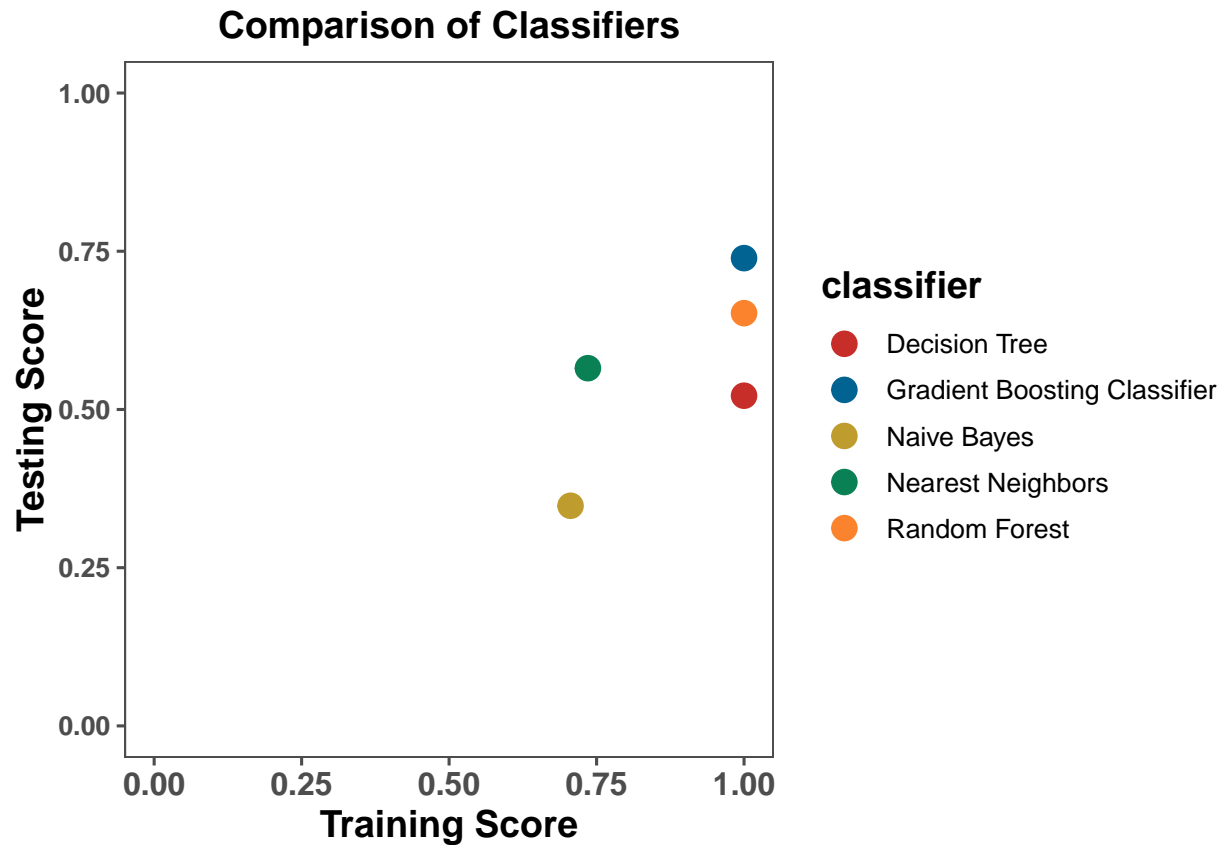
clf_compare = read.csv("classifier_compare.csv", header = TRUE, sep = ",")

clf_compare = clf_compare %>% arrange(desc(test_score))

clf_compare = clf_compare %>% filter(!classifier %in% c("Logistic Regression",
                                                       "Linear SVM",
                                                       "Neural Net"))

clf_compare %>%
  ggplot(aes(x = train_score, y = test_score)) +
  geom_point(aes(color = classifier), size = 4) +
  theme_few() +
  xlim(0,1) +
  ylim(0,1) +
  scale_color_wsj() +
  xlab("Training Score") + ylab("Testing Score") +
  ggtitle("Comparison of Classifiers") +
  theme(axis.text = element_text(face = "bold", size = rel(0.8)),
        axis.title = element_text(face = "bold", size = rel(1.2)),
        legend.title = element_text(face = "bold", size = rel(1.2)),
        axis.text.x = element_text(face = "bold", size = rel(1.2)),
        plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))

```



Lexical Diversity and Length

```
ld = read.csv("lexical_diversity.csv", header = TRUE, sep = ",")

ld$R = ifelse(ld$Replicated == "yes", 1,0)
ld$ld_c = scale(ld$Lexical.Diversity, center = TRUE, scale = FALSE)
ld_lm = glm(data = ld, R ~ Ldnew + Length,
             family = "binomial")
summary(ld_lm)
```

```
##
## Call:
## glm(formula = R ~ Ldnew + Length, family = "binomial", data = ld)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3629  -0.9831  -0.7739   1.2071   1.8834
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.820672   3.104206  -0.587   0.5575
## Ldnew       -1.640713   3.339191  -0.491   0.6232
## Length       0.015536   0.007623   2.038   0.0415 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```

##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 122.16  on 90  degrees of freedom
## Residual deviance: 115.76  on 88  degrees of freedom
## AIC: 121.76
##
## Number of Fisher Scoring iterations: 4
ld_lm2 = glm (data = ld, R ~ Length,
              family = "binomial")
summary(ld_lm2)

##
## Call:
## glm(formula = R ~ Length, family = "binomial", data = ld)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3662  -0.9786  -0.7906   1.2452   1.8789
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -3.23329     1.21859  -2.653  0.00797 **
## Length       0.01690     0.00716   2.360  0.01828 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 122.16  on 90  degrees of freedom
## Residual deviance: 116.00  on 89  degrees of freedom
## AIC: 120
##
## Number of Fisher Scoring iterations: 4
anova(ld_lm, ld_lm2)

## Analysis of Deviance Table
##
## Model 1: R ~ Ldnew + Length
## Model 2: R ~ Length
##   Resid. Df Resid. Dev Df Deviance
## 1          88      115.76
## 2          89      116.00 -1  -0.24222
## need to plot length figure

x = sjPlot::plot_model(ld_lm, type = "pred", terms = "Length")

x + theme_few() +
  scale_color_ws() +
  xlab("Length of Abstract") + ylab("Predicted probability of Replication") +
  ggtitle("Length of Abstract Predicting Replication") +
  theme(axis.text = element_text( face = "bold", size = rel(0.8)),
        axis.title = element_text(face = "bold", size = rel(1.2)),

```

```

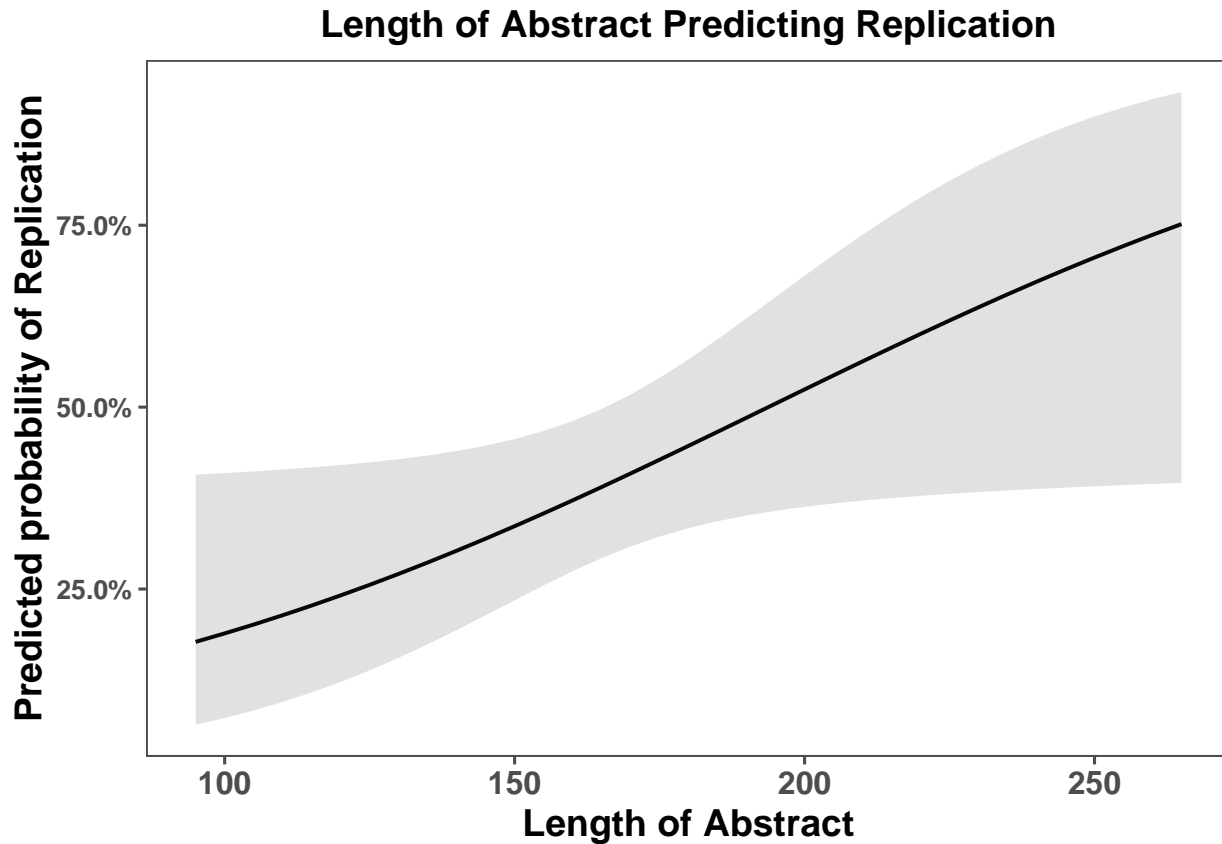
legend.title = element_text(face = "bold", size = rel(1.2)),
axis.text.x = element_text(face = "bold", size = rel(1.2)),
plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))

```

```

## Scale for 'colour' is already present. Adding another scale for
## 'colour', which will replace the existing scale.

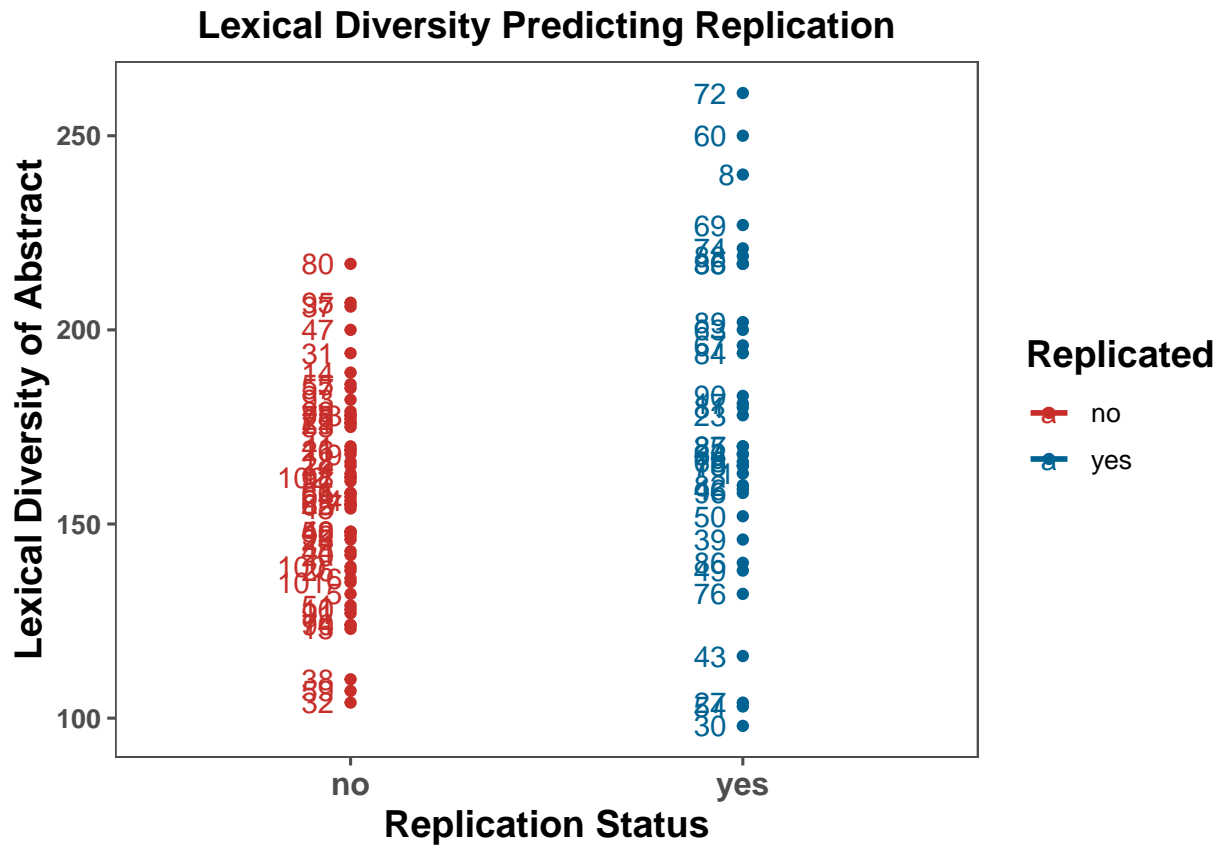
```



```

ld$Lexical.Diversity = round(ld$Lexical.Diversity, digits = 2)
ld %>%
  ggplot(aes(x = Replicated, y = Length, color = Replicated)) +
    geom_point() +
    geom_text(aes(label = TextNumber), hjust = 1.5, vjust = .5) +
    geom_smooth(method = "glm", se = FALSE) +
    theme_few() +
    scale_color_wsj() +
    xlab("Replication Status") + ylab("Lexical Diversity of Abstract") +
    ggtitle("Lexical Diversity Predicting Replication") +
    theme(axis.text = element_text(face = "bold", size = rel(0.8)),
          axis.title = element_text(face = "bold", size = rel(1.2)),
          legend.title = element_text(face = "bold", size = rel(1.2)),
          axis.text.x = element_text(face = "bold", size = rel(1.2)),
          plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))

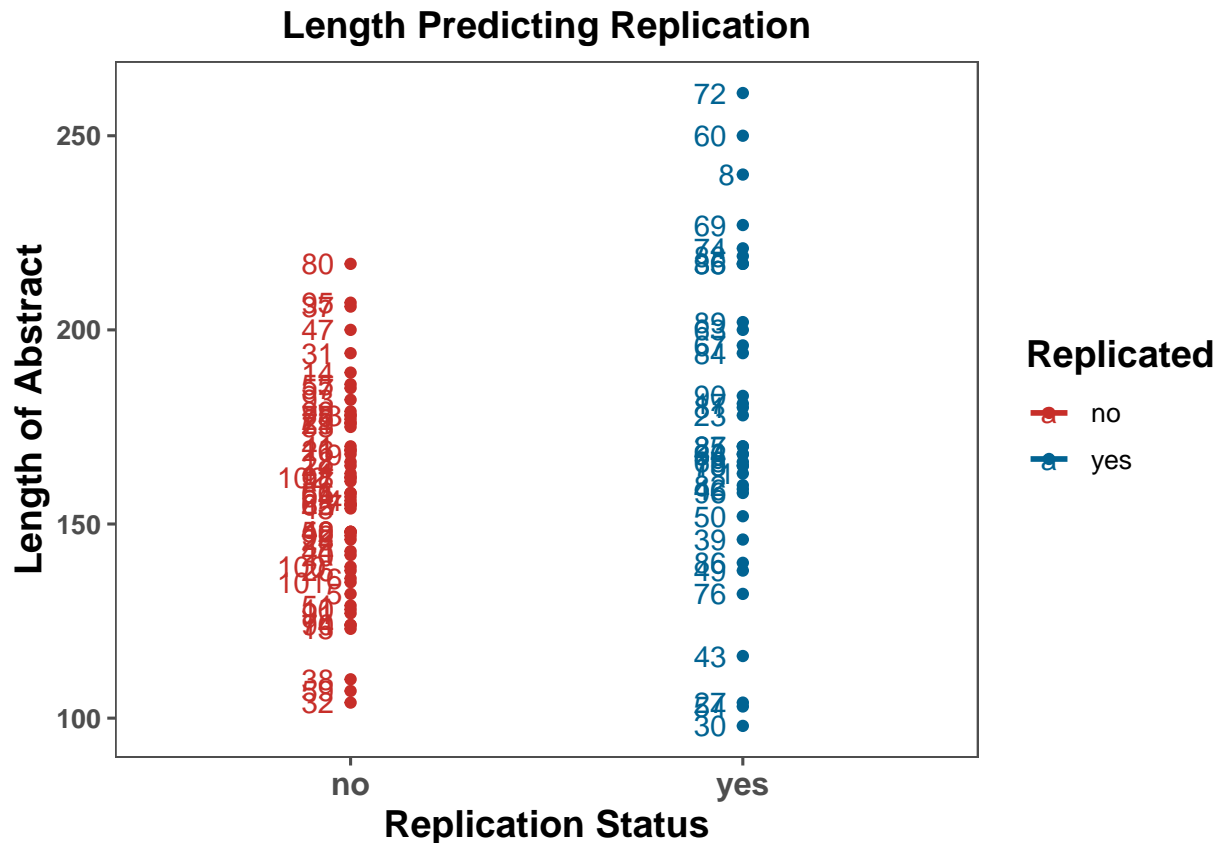
```



```

1d %>%
  ggplot(aes(x =Replicated, y = Length, color = Replicated)) +
    geom_point()+
    geom_text(aes(label=TextNumber), hjust = 1.5, vjust = .5)+
    geom_smooth(method = "glm", se = FALSE)+
    theme_few()+
    scale_color_wsj()+
    xlab("Replication Status") + ylab("Length of Abstract") +
    ggtitle("Length Predicting Replication") +
    theme(axis.text = element_text( face = "bold", size = rel(0.8)),
          axis.title = element_text(face = "bold", size = rel(1.2)),
          legend.title = element_text(face = "bold", size = rel(1.2)),
          axis.text.x = element_text(face = "bold", size = rel(1.2)),
          plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))

```



POS Tagging

```
pos_data = read.csv("pos_python.csv", header = TRUE, sep = ",")
library(dplyr)
## This data is in wide format: need to convert to long format

pos_long = tidyr::gather(pos_data, PartOfSpeech, Count,
                          Adjective, Noun, Verb, Other, factor_key=TRUE)
pos_long = pos_long %>% arrange(TextNumber)

pos_long$Percent = pos_long$Count/pos_long$Length
pos_long$Percent = round(pos_long$Percent, digits = 2)

pos_long$R = ifelse(pos_long$Replicated == "yes", 1,0)

contrasts(pos_long$PartOfSpeech) = contr.treatment(4, base = 1)
library(lme4)

## Warning: package 'lme4' was built under R version 3.4.4
## Loading required package: Matrix
## Warning: package 'Matrix' was built under R version 3.4.4

c13 <- glmerControl(optimizer="optimx",
                    optCtrl=list(method="nlopt",maxiter=10000))
```



```

library(optimx)
pos_lm = glmer (data = pos_long, R ~ PartOfSpeech*Percent +
                (1|TextNumber),
                family = "binomial", control = cl3)
summary(pos_lm)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: R ~ PartOfSpeech * Percent + (1 | TextNumber)
## Data: pos_long
## Control: cl3
##
##      AIC      BIC   logLik deviance df.resid
##    132.7    167.8    -57.3    114.7     355
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -0.002221 -0.001736 -0.001580  0.036242  0.042923
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## TextNumber (Intercept) 3457     58.8
## Number of obs: 364, groups: TextNumber, 91
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -12.5772     6.1014  -2.061  0.0393 *
## PartOfSpeech2     0.9861    14.0685   0.070  0.9441
## PartOfSpeech3     0.4694     6.8714   0.068  0.9455
## PartOfSpeech4    -1.1406     9.6022  -0.119  0.9054
## Percent          -0.4872    29.6337  -0.016  0.9869
## PartOfSpeech2:Percent -2.0591    42.2061  -0.049  0.9611
## PartOfSpeech3:Percent -6.2600    49.8876  -0.125  0.9001
## PartOfSpeech4:Percent  4.0429    39.6274   0.102  0.9187
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) PrtOS2 PrtOS3 PrtOS4 Percnt POS2:P POS3:P
## PartOfSpch2 -0.409
## PartOfSpch3 -0.826  0.373
## PartOfSpch4 -0.695  0.246  0.580
## Percent     -0.945  0.412  0.828  0.697
## PrtOfSpc2:P  0.664 -0.929 -0.593 -0.447 -0.705
## PrtOfSpc3:P  0.544 -0.260 -0.872 -0.356 -0.579  0.425
## PrtOfSpc4:P  0.783 -0.296 -0.663 -0.963 -0.826  0.547  0.440
car::Anova(pos_lm)

## Analysis of Deviance Table (Type II Wald chisquare tests)
##
## Response: R
##
##              Chisq Df Pr(>Chisq)

```

```
## PartOfSpeech      0.0001  3      1.0000
## Percent           0.0000  1      0.9964
## PartOfSpeech:Percent 0.0530  3      0.9968

pos_long$Replicated = ifelse(pos_long$Replicated == "yes", "Yes", "No")
```

```
library(ggplot2)
library(ggthemes)
pos_long %>%
  ggplot(aes(x =Replicated, y = Percent, color = Replicated)) +
    geom_point()+
    geom_text(aes(label=TextNumber), hjust = 1.5, vjust = .5)+
    geom_smooth(method = "glm", se = FALSE)+
    theme_light()+
    scale_color_wsj()+
    facet_wrap(~PartOfSpeech)+
    xlab("Replication Status") + ylab("Percentage of POS") +
    ggtitle("Parts of Speech Predicting Replication?" ) +
    theme(axis.text = element_text( face = "bold", size = rel(0.8)),
          axis.title = element_text(face = "bold", size = rel(1.2)),
          legend.title = element_text(face = "bold", size = rel(1.2)),
          axis.text.x = element_text(face = "bold", size = rel(1.4)),
          strip.text.x = element_text(face = "bold", size = rel(1.2)),
          plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))
```

