The Influence of Text Characteristics for Email Classification

Group 22

```
library(ggplot2)
library(dplyr)
library(moderndive)
library(gapminder)
library(skimr)
library(tidyverse)
library(gt)
library(gtchwork)
library(gridExtra)
library(broom)
library(knitr)
library(GGally)
```

email<-read.csv("C:/Users/70652/Desktop/STATS5085 Data Analysis Skills/Project 2/DAS-Group-2

```
email$yesno<-as.factor(email$yesno)</pre>
```

1 Exploratory Data Analysis

1.1 Correlation

```
ggpairs(email[,1:6]) +
  theme(plot.background = element_rect(
    fill = "transparent",
    colour = NA,
    size = 1))
```

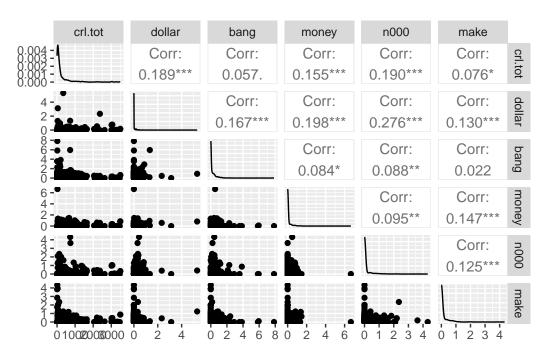


Figure 1: Correlations between each variables.

1.2 Data Visualization

Spam indictor with total length of uninterrupted sequences of

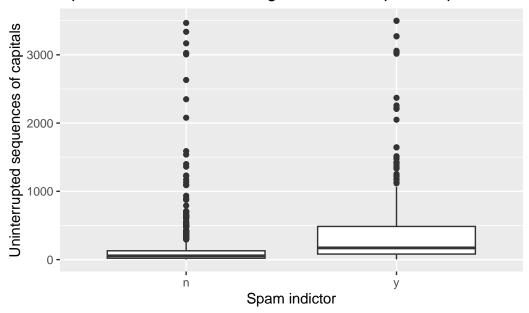


Figure 2: Boxplot of total length of uninterrupted sequences of capitals.

Spam indictor with occurrences of the dollar sign

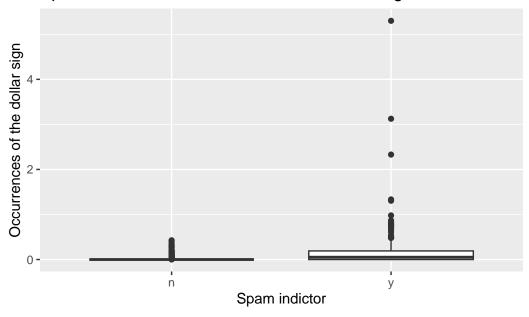


Figure 3: Boxplot of occurrences of the dollar sign.

Spam indictor with occurrences of "!"

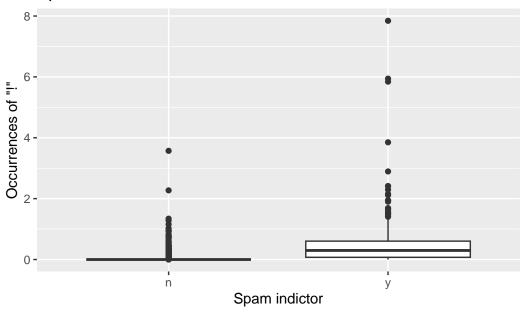


Figure 4: Boxplot of occurrences of '!'.

Spam indictor with occurrences of "money"

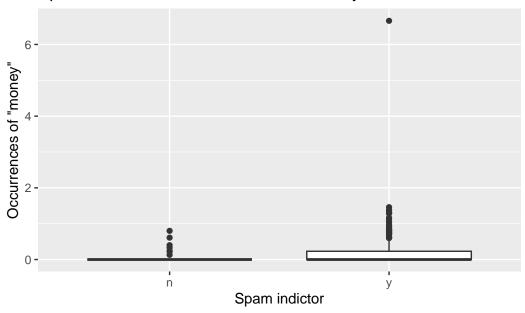


Figure 5: Boxplot of occurrences of "money".

```
ggplot(email, aes(x = yesno, y = n000)) +
  geom_boxplot() +
  labs(x = "Spam indictor", y = 'Occurrences of "000"',
     title = 'Spam indictor with occurrences of "000"')
```

Spam indictor with occurrences of "000"

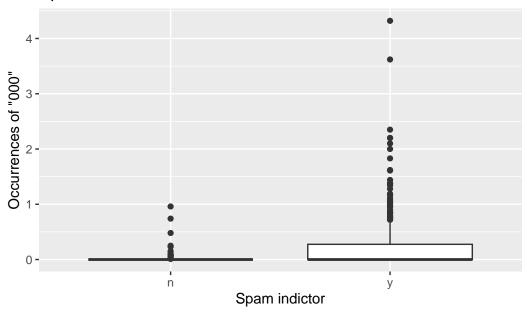


Figure 6: Boxplot of occurrences of '000'.

```
ggplot(email, aes(x = yesno, y = make)) +
  geom_boxplot() +
  labs(x = "Spam indictor", y = 'Occurrences of "make"',
       title = 'Spam indictor with occurrences of "make"')
```

Spam indictor with occurrences of "make"

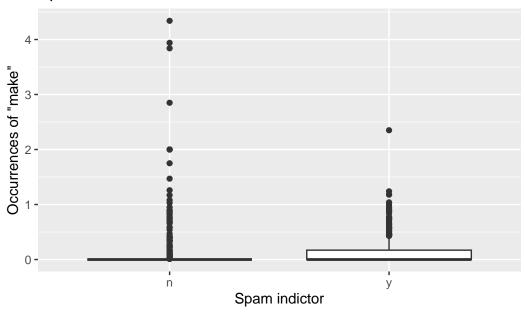


Figure 7: Boxplot of occurrences of 'make'.

2 Formal Data Analysis

```
Call:
```

```
glm(formula = yesno ~ crl.tot + dollar + bang + money + n000 +
    make, family = binomial(link = "logit"), data = email)
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|) (Intercept) -1.8101190 0.1254636 -14.427 < 2e-16 *** crl.tot 0.0005502 0.0001886 2.917 0.003533 ** dollar 8.1346140 1.5396484 5.283 1.27e-07 *** bang 2.9172085 0.3363971 8.672 < 2e-16 ***
```

```
5.9724851 1.2455257 4.795 1.63e-06 ***
money
n000
            3.4827736 1.0261134 3.394 0.000688 ***
make
           -0.4553154  0.4065463  -1.120  0.262731
___
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 1234.17 on 919 degrees of freedom
Residual deviance: 752.44 on 913 degrees of freedom
AIC: 766.44
Number of Fisher Scoring iterations: 7
mod1coefs <- round(coef(model1), 2)</pre>
model2 <- glm(yesno ~ crl.tot+dollar+bang+money+n000, data = email,</pre>
            family = binomial(link = "logit"))
summary(model2)
Call:
glm(formula = yesno ~ crl.tot + dollar + bang + money + n000,
    family = binomial(link = "logit"), data = email)
Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.8455026 0.1229712 -15.008 < 2e-16 ***
           0.0005579 0.0001887 2.956 0.003115 **
crl.tot
dollar
            8.1812828 1.5395890 5.314 1.07e-07 ***
            2.9348590 0.3371484 8.705 < 2e-16 ***
bang
            5.8334954 1.2421522 4.696 2.65e-06 ***
money
n000
            3.4273127 1.0224981 3.352 0.000803 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 1234.17 on 919 degrees of freedom
Residual deviance: 754.04 on 914 degrees of freedom
```

AIC: 766.04

Number of Fisher Scoring iterations: 7

mod2coefs <- round(coef(model2), 3)</pre>

$$\begin{split} \ln\left(\frac{p}{1-p}\right) &= \alpha + \beta_{crl.tot} \cdot \text{crl.tot} + \beta_{dollar} \cdot \text{dollar} + \beta_{bang} \cdot \text{bang} + \\ &\beta_{money} \cdot \text{money} + \beta_{n000} \cdot \text{n000} \\ &= -1.846 + 0.001 \cdot \text{crl.tot} + 8.181 \cdot \text{dollar} + 2.935 \cdot \text{bang} + 5.833 \cdot \text{money} + 3.427 \cdot \text{n000} \end{split}$$

confint(model2) %>%
 kable()

	2.5~%	97.5 %
(Intercept)	-2.0926399	-1.610104
crl.tot	0.0001855	0.000936
dollar	5.3097543	11.355237
bang	2.3039250	3.626788
money	3.6506969	8.565015
n000	1.6521512	5.709387