# Thesis Data Analysis: The Psychology of Scientific Fraud

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# Introduction

This document is intended to track the data analysis for my undergraduate thesis project on the psychology of scientific fraud. I have already planned out (and preregistered) my data analysis plan (see here), and I will note throughout the document if I deviate from this plan and why. This project is also being tracked in a private GitHub repository in case I need to revert back to a previous version of the project due to a fatal error.

# **Overview of Sections**

- Data Cleaning: The data cleaning section will take the finalized dataset that I have included in the project directory and use it to create the remaining variables that we need for our analysis. We will also save this dataset.
- Data Exploration: In the data exploration section, we will check for and deal with missing data, run descriptive statistics of our outcome variables, visualize our data, and run bivariate correlations.
- **Testing Hypotheses:** In the hypothesis testing section, we will test each of our hypotheses one-by-one, according to our analysis plan.

# **Directory Set-Up**

In order to ensure that this analysis is computationally reproducible, I have included everything that is needed to complete this analysis (just the finalized data file, "study\_dataset.csv") in the current working directory. The code will be displayed before the output of each analysis.

#### Variable Definitions

Although the following variables may not exist until after the data cleaning section, here is what each variable name refers to, so that you can refer to them while examining the code.

DOI: Unique identifier for each paper (i.e., the paper DOI).

PaperType: Categorical variable indicating SAFP, SAGP, MAFP, or MAGP.

LingObf: Continuous variable for linguistic obfuscation.

CertSent: Continuous variable for certainty sentiment.

Refs: Count variable for references.

FraudCorrAuth: Dichotomous variable indicating if the fraudulent author is the corresponding author (1) or is not (0). Unknown cases will be marked in a seperate variable with this variable left blank.

NumAuth: Count variable indicating the number of authors for each paper.

abstraction: Abstraction index composed of the sum of standardized scores for article, prep, and quantity.

article: Articles from LIWC.

prep: Prepositions from LIWC.

quantity: Quantities from LIWC.

cause: Causation terms from LIWC.

jargon: The percent of words not captured by LIWC (100-Dic).

Dic: The percentage of words captured by all LIWC dictionaries.

emo\_pos: Positive emotion terms from LIWC.

flesch\_re: Flesch Reading Ease from ARTE.

## **Initial Package Installations**

If you would like to reproduce this analysis, here are the package I will be using, so that they can be cued before starting.

```
install.packages("readr")  # For reading data
```

Installing package into '/Users/benjaminzubaly/Library/R/x86\_64/4.3/library' (as 'lib' is unspecified)

The downloaded binary packages are in /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded\_packages

```
install.packages("dplyr")  # For data manipulation and handling of missing data
```

Installing package into '/Users/benjaminzubaly/Library/R/x86\_64/4.3/library' (as 'lib' is unspecified)

The downloaded binary packages are in /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded\_packages

```
install.packages("psych") # For descriptive statistics, correlations
```

Installing package into '/Users/benjaminzubaly/Library/R/x86\_64/4.3/library' (as 'lib' is unspecified)

The downloaded binary packages are in /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded\_packages

```
install.packages("ggplot2") # For data visualization
Installing package into '/Users/benjaminzubaly/Library/R/x86_64/4.3/library'
(as 'lib' is unspecified)
The downloaded binary packages are in
    /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded_packages
  install.packages("car")
                                 # For diagnostic tests such as Levene's test
Installing package into '/Users/benjaminzubaly/Library/R/x86_64/4.3/library'
(as 'lib' is unspecified)
The downloaded binary packages are in
    /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded_packages
  install.packages("rcompanion") # For Games-Howell post-hoc test (if applicable)
Installing package into '/Users/benjaminzubaly/Library/R/x86_64/4.3/library'
(as 'lib' is unspecified)
The downloaded binary packages are in
    /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded_packages
  install.packages("dunn.test") # For Dunn post-hoc test (if applicable)
Installing package into '/Users/benjaminzubaly/Library/R/x86_64/4.3/library'
(as 'lib' is unspecified)
The downloaded binary packages are in
    /var/folders/38/1ybnplc53zdb089bn6drqfn00000gn/T//Rtmp94Cdsa/downloaded_packages
```

# **Initial Import of Data**

We will not load in the dataset "study\_dataset.csv" from the working directory.

```
library(readr) # Loading the readr package

data <- read_csv("study_dataset.csv") # Loading in study dataset as "data"

Rows: 88 Columns: 207
-- Column specification -------
Delimiter: ","
chr (26): DOI, Title, Subject, Institution, Journal, Publisher, Country, Au...
dbl (181): Record ID, RetractionPubMedID, OriginalPaperDate, year, OriginalP...
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.</pre>
```

I have viewed the data frame, and the data seems to have loaded correctly.

# **Data Cleaning**

To conduct the analysis, we will first need to calculate the LingObf variable by calculating the abstraction index and jargon words; creating standardized scores for abstraction, cause, jargon, emo\_pos, and flesch\_re; and calculating the LingObf composite variable from these standardized scores.

1. First, we will calculate the abstraction index by creating standardized scores for article, prep, and quantity and summing them.

```
# Calculate standardized scores for article, prep, and quantity and add them to the datase
data$articles_standardized <- scale(data$article)
data$prep_standardized <- scale(data$prep)
data$quantity_standardized <- scale(data$quantity)

# Create the new variable 'abstraction' as the sum of the three standardized variables
data$abstraction <- (data$articles_standardized + data$prep_standardized + data$quantity_s</pre>
```

- After viewing the data, the transformations and variable calculation seem to have occurred appropriately.
- 2. Next, we will calculate the jargon words by subtracting Dic from 100.

```
# Calculate the new variable 'jargon' by subtracting 'Dic' from 100
data$jargon <- (100 - data$Dic)</pre>
```

- After viewing the data, the variable calculation seem to have occurred appropriately.
- 3. Next, we will create standardized scores for each subcomponent of the LingObf.

```
# Standardize the new set of variables and add them to the dataset
data$abstraction_standardized <- scale(data$abstraction)
data$cause_standardized <- scale(data$cause)
data$jargon_standardized <- scale(data$jargon)
data$emo_pos_standardized <- scale(data$emo_pos)
data$flesch_re_standardized <- scale(data$flesch_re)</pre>
```

- After viewing the data, the variable transformations seem to have occurred appropriately.
- 4. Now we will calculate the LingObf variable using the following formula: [cause\_standardized + abstraction\_standardized + jargon\_standardized] [emo\_pos\_standardized + flesch re standardized].

```
# Calculate 'LingObf'
data$LingObf <- (data$cause_standardized + data$abstraction_standardized + data$jargon_standardized +
```

- After viewing the data, the variable calculation seem to have occurred appropriately.
- 5. Lastly, our variable that indicates certainty sentiment is currently certainty\_avg, but to make things easier I am going to copy this data into a new variables called CertSent.

```
data$CertSent <- data$certainty_avg</pre>
```

To ensure that our clean data is saved, we will write the dataset to the current working directory.

```
# Writing our data as a csv file in the current working directory
write.csv(data, "clean_study_data.csv")
```

• I have opened the saved data file outside of Rstudio, and it seems to have been written correctly.

# **Data Exploration**

#### **Dealing with Missing Data**

1. Data will be first inspected for missing scores.

```
library(dplyr) # Loading the dplyr package for data manipulation and handling missing value.

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':
```

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

filter, lag

```
# To summarize the number of missing values in each column
missing_data_summary <- sapply(data, function(x) sum(is.na(x)))
print(missing_data_summary) # To see summary of missing values for all columns</pre>
```

D 1 TD	DOT
Record ID	DOI
44	0
Subject	Title
44	0
Journal	Institution
0	0
Country	Publisher
0	0
URLS	Author
68	0
RetractionDate	ArticleType
44	0
${ t RetractionPubMedID}$	RetractionDOI
51	48
year	${\tt OriginalPaperDate}$
0	44
${\tt OriginalPaperPubMedID}$	OriginalPaperDOI
51	0
Reason	RetractionNature
44	44
Notes	Paywalled
70	44

simple_reason	NumAuth
44	0
inst_pres	PaperType
0	0
gender	matched_DOI_SAFP_MAFP
0	44
matched_DOI_SAFP_SAGP	matched_DOI_MAFP_MAGP
44	44
matching_concessions	FraudCorrAuth
42	80
FCA_unknown	FCA_notes
66	66
different_country	year_difference
22	22
different_gender	different_inst_pres
22	22
different_journal	Refs
66	0
Abstract	flesch_re
0	0
wordcount	valence_min
0	0
valence_max	valence_avg
0	0
valence_std 0	extremity_min O
extremity_max 0	extremity_avg 0
extremity_std	extremity_min_pos
extremity_std	extremity_min_pos
extremity_max_pos	extremity_avg_pos
0	0.010m10y_avg_pob
extremity_std_pos	extremity_min_neg
0	1
extremity_max_neg	extremity_avg_neg
1	1
extremity_std_neg	extremity_PosMinNeg
5	0
emotionality_min	emotionality_max
0	0
emotionality_avg	emotionality_std
0	0
<pre>emotionality_min_pos</pre>	${\tt emotionality\_max\_pos}$

0	0
emotionality_avg_pos	${\tt emotionality\_std\_pos}$
0	0
<pre>emotionality_min_neg 1</pre>	<pre>emotionality_max_neg 1</pre>
<pre>emotionality_avg_neg 1</pre>	emotionality_std_neg 5
$\begin{array}{c} \texttt{emotionality\_PosMinNeg} \\ \texttt{0} \end{array}$	<pre>count_unique_evaluative_pos</pre>
<pre>count_total_evaluative_pos 0</pre>	<pre>count_unique_evaluative_neg 0</pre>
<pre>count_total_evaluative_neg</pre>	<pre>count_unique_evaluative 0</pre>
<pre>count_total_evaluative 0</pre>	count_evaluative_PosMinNeg 0
ambivalent 0	pos_dichotomous 87
certainty_min	certainty_max
0	0
certainty_avg	certainty_std
0	0
count_unique_certainty 0	<pre>count_total_certainty 0</pre>
Segment	WC
0	0
Analytic	Clout
0	0
Authentic	Tone
0	0
WPS	BigWords
0	0
Dic	Linguistic
0	0
function	pronoun
0	0
ppron	i
0	0
we O	you 0
shehe	they
shene 0	O
ipron	det
0	0
ŭ	0

article	number
0	0
prep	auxverb
0	0
adverb	conj
0	0
negate	verb
0	0
adj	quantity
0	0
Drives	affiliation
0	0
achieve	power
0	0
Cognition	allnone
0	0
cogproc	insight
0	0
cause	discrep
0	0
tentat	certitude
0	0
differ	memory
0	0
Affect	tone_pos
0	0
tone_neg	emotion
0	0
emo_pos	emo_neg
0	0
emo_anx	emo_anger
0	0
emo_sad	swear
0	0
Social	socbehav
0	0
prosocial	polite
0	0
conflict	moral
0	0
comm	socrefs
0	0
family	friend

0	0
male	female
0	0
politic	Culture
0	0
tech	ethnicity
0	0
leisure	Lifestyle
0	0
work	home
0	0
relig	money
0	0
health	Physical
0	0
wellness	illness
0	0
substances	mental
0	0
food	sexual
0	0
need	death
0	0
acquire	want
0	0
fulfill	lack
0	0
reward	fatigue
0	0
curiosity	risk
0	0
Perception	allure
0	0
motion	attention
0	0
visual	space
0	0
feeling	auditory
focusars at	0
focuspast	time
focusfuture	0
focusfuture	focuspresent
0	0

```
Conversation
                                              netspeak
                                                 nonflu
                   assent
                                                      0
                   filler
                                                AllPunc
                                                      0
                   Period
                                                  Comma
                    QMark
                                                 Exclam
                        0
                                                      0
                                                 OtherP
                  Apostro
                        0
                                                      0
                    Emoji
                                 articles_standardized
                        0
                                                      0
       prep_standardized
                                 quantity_standardized
             abstraction
                                                 jargon
abstraction_standardized
                                    cause_standardized
     jargon_standardized
                                  emo_pos_standardized
                                               LingObf
 flesch_re_standardized
                 CertSent
                        0
```

• Taking a look at our outcome variables, there are no missing scores, so we do not need to impute any values.

#### **Descriptive Statistics**

1. I will now generate descriptive statistics for each relevant variable in the dataset. I originally (in the preregistered plan) was simply going to deploy the describe() function on the entire dataset, but because I retained all of the columns from the Retraction Watch Database (Retraction Watch Database, 2023) and all of the output from the text analysis packages (Aggarwal, 2022; Boyd et al., 2022; Rocklage et al., 2023) there are currently 219 variables. Because this would be unmanageable, I am going to only calculate descriptive statistics for a selection of variables of interest. I will first create a dataframe with only the continuous variables that I am interested in generating descriptive statistics for, and I will use the psych package to produce the descriptive statistics for these variables.

# library(psych) # Loading the psych package

- # Selecting the continuous variables I am interested in getting descriptive statistics for continuous\_data\_for\_descriptives <- data[c("year", "Refs", "flesch\_re", "WC", "abstraction")
- # Generating descriptive statistics for variables of interest
  continuous\_descriptive\_stats\_all <- describe(continuous\_data\_for\_descriptives)</pre>
- # Displaying the results of the descriptive stats for the variables listed above
  continuous\_descriptive\_stats\_all

	vars	n	mean	sd	median	trimmed	mad	min	max
year			2009.70		2013.00			1980.00	2022.00
Refs		88	45.82	23.47	39.00	43.21	20.02	12.00	146.00
flesch_re		88	34.90	10.94	36.45	35.46		-10.89	56.59
WC -									13901.00
abstraction	5	88	0.00	1.83	0.09	0.04	1.86	-4.46	3.55
jargon	6	88	33.33	7.47	34.76	33.50	7.84	17.91	46.67
CertSent	7	88	6.17	0.25	6.18	6.18	0.22	5.21	6.76
LingObf	8	88	0.00	1.67	-0.14	-0.01	1.54	-5.84	4.28
cause	9	88	2.80	1.22	2.62	2.69	1.02	0.89	6.01
emo_pos	10	88	0.09	0.22	0.04	0.05	0.06	0.00	1.67
article	11	88	8.38	2.20	8.31	8.33	2.13	3.53	14.47
prep	12	88	15.43	1.68	15.66	15.51	1.34	9.76	20.67
quantity	13	88	4.23	1.32	4.08	4.14	1.33	1.69	8.66
PaperType*	14	88	2.50	1.12	2.50	2.50	1.48	1.00	4.00
	ra	ange	e skew k	kurtosis	se				
year	42	2.00	-1.25	0.74	1.12				
Refs	134	1.00	1.41	2.87	2.50				
flesch_re	67	7.48	3 -0.85	2.16	1.17				
WC	12777	7.00	1.21	2.18	239.66				
abstraction	8	3.01	L -0.20	-0.42	0.20				
jargon	28	3.76	6 -0.27	-0.94	0.80				
CertSent		1.55	5 -0.69	1.42	0.03				
LingObf	10	0.12	2 -0.21	1.40	0.18				
cause		5.12	0.80	0.14	0.13				
emo_pos		1.67	7 5.66	34.22	0.02				
article	10	).94	1 0.33	-0.19	0.23				
prep	10	0.91	L -0.44	1.22	0.18				
quantity	6	3.97	7 0.67	0.42	0.14				
PaperType*	3	3.00	0.00	-1.40	0.12				

- I won't make too many comments here, because for most of these measures there are no formal or informal norms against which to judge them. That being said:
  - Year: The mean year is mid-2009, with a standard deviation of 10 years (max 2022 and min 1980), indicating that the sample is recent enough to be relevant but also spans quite a number of years. This I think is good insofar as the recent history of academic publishing is represented more fully (some recent investigations limited their search to the three years prior to publication). There is some negative skew, which I suspect is due to the 1980 paper being quite a bit older than most papers.
  - Refs: The mean number of references is 45.82 with a standard deviation of 23.47. At least intuitively, this seems like a pretty standard distribution of references if we were to randomly select papers from the literature. However, the range is huge, with one paper showing 146 references—perhaps why the skew is positive. This may be something to consider when making group comparisons, as this point may have significant leverage.
  - WC: The average word count is 4436.25, and the standard deviation is 2248.24. It is also positively skewed, and although this should not affect our linguistic dependent variables it may be a confounding factor for references. Indeed, taking a look at the dataset I can see that the second highest value for word count (9670) also has the maximum value for references (the outlier noted above at 146 references). This will be something to watch out for.
- Now we will create descriptive statistics for each of the continuous variables above within the PaperType groups.

```
# Making PaperType a factor variable in the continuous variable dataframe to allow for gro
continuous_data_for_descriptives$PaperType <- factor(continuous_data_for_descriptives$Paper
# Generating descriptive statistics within PaperType groups
descriptive_stats_by_PaperType <- describeBy(continuous_data_for_descriptives, group = condescriptive_stats_by_PaperType</pre>
```

Descriptive statistics by group group: Single-Authored Fraudulent Papers

	vars	n	mean	sd	median	trimmed	mad	min	max
year	1	22	2008.95	11.74	2013.00	2010.39	6.67	1983.00	2022.00
Refs	2	22	44.18	19.93	36.00	42.89	17.79	18.00	87.00
flesch_re	3	22	36.09	9.17	37.06	36.29	10.53	18.49	54.36
WC	4	22	3998.91	1741.31	3796.50	3911.39	1829.53	1273.00	7330.00
abstraction	5	22	0.21	1.54	0.04	0.27	1.25	-3.27	2.95
jargon	6	22	34.23	8.33	36.94	34.90	8.49	17.91	44.35
CertSent	7	22	6.21	0.21	6.18	6.20	0.18	5.87	6.66

LingObf	8 22	-0.24	1.79	-0.04	-0.05	1.97	-4.23	2.25
cause	9 22	2.71	1.48	2.50	2.56	1.07	0.89	6.01
emo_pos	10 22	0.15	0.35	0.06	0.08	0.06	0.00	1.67
article	11 22	8.35	2.60	8.48	8.41	2.90	3.53	12.25
prep	12 22	15.75	1.60	15.75	15.79	1.43	11.79	18.42
quantity	13 22	4.27	0.97	4.11	4.21	1.13	2.59	6.59
PaperType	14 22	1.00	0.00	1.00	1.00	0.00	1.00	1.00
	range	skew ku	ırtosis	se				
year	39.00	-1.02	-0.29	2.50				
Refs	69.00	0.56	-1.06	4.25				
flesch_re	35.87	-0.01	-0.63	1.95				
WC	6057.00	0.54	-0.67 3	371.25				
abstraction	6.21	-0.23	-0.33	0.33				
jargon	26.44	-0.56	-1.04	1.78				
CertSent	0.78	0.44	-0.57	0.04				
LingObf	6.48	-0.70	-0.32	0.38				
cause	5.12	0.93	-0.05	0.32				
emo_pos	1.67	3.81	13.82	0.07				
article	8.72	-0.16	-1.18	0.55				
prep	6.63	-0.34	-0.15	0.34				
quantity	4.00	0.49	-0.41	0.21				
${\tt PaperType}$	0.00	NaN	NaN	0.00				

group: Multi-Authored Fraudulent Papers

	vars	n	mean	sd	median	trimmed	mad	min	max
year	1	22	2010.14	9.59	2012.50	2011.67	8.15	1982.00	2022.00
Refs	2	22	44.00	19.66	39.50	42.06	17.05	18.00	94.00
flesch_re	3	22	34.18	9.64	31.41	33.58	9.38	20.08	56.59
WC	4	22	4705.91	2047.63	4274.00	4559.39	1931.83	1823.00	9004.00
abstraction	5	22	-0.45	1.83	-0.53	-0.44	1.95	-4.10	3.55
jargon	6	22	35.97	6.97	37.36	36.23	6.28	23.85	46.48
CertSent	7	22	6.22	0.18	6.20	6.23	0.24	5.82	6.47
LingObf	8	22	0.47	1.43	0.18	0.36	1.57	-1.66	4.13
cause	9	22	2.84	1.08	2.74	2.75	1.03	1.14	5.66
emo_pos	10	22	0.03	0.04	0.01	0.02	0.02	0.00	0.15
article	11	22	8.19	1.94	8.51	8.12	1.53	4.90	12.66
prep	12	22	14.84	1.64	15.19	14.84	1.75	11.80	17.58
quantity	13	22	4.22	1.39	4.10	4.15	1.54	2.27	6.95
${\tt PaperType}$	14	22	2.00	0.00	2.00	2.00	0.00	2.00	2.00
	rai	nge	skew ku	ırtosis	se				
year	40	.00	-1.37	1.52	2.05				
Refs	76	.00	0.86	-0.01	4.19				
flesch re	36	.51	0.59	-0.62	2.05				

WC	7181.00	0.56	-0.81	436.56
${\tt abstraction}$	7.65	0.02	-0.64	0.39
jargon	22.63	-0.38	-1.16	1.49
CertSent	0.64	-0.27	-1.00	0.04
LingObf	5.79	0.68	-0.15	0.30
cause	4.52	0.78	0.30	0.23
emo_pos	0.15	1.55	1.39	0.01
article	7.76	0.16	-0.34	0.41
prep	5.78	-0.11	-1.18	0.35
quantity	4.68	0.46	-1.11	0.30
PaperType	0.00	NaN	NaN	0.00

group: Single-Authored Genuine Pape:
--------------------------------------

group: Single-Authored Genuine Papers								
	vars n	mean	sd	median	trimmed	mad	min	max
year	1 22	2008.95	12.24	2013.50	2010.61	5.93	1980.00	2022.00
Refs	2 22	51.05	32.53	40.50	46.28	28.91	12.00	146.00
flesch_re	3 22	34.46	15.72	38.40	36.15	10.34	-10.89	54.97
WC	4 22	4632.36	2424.81	4148.00	4491.00	2364.01	1141.00	9670.00
abstraction	5 22	0.80	1.89	1.00	0.91	2.13	-3.39	3.43
jargon	6 22	29.50	6.19	31.79	29.61	5.77	19.26	38.66
CertSent	7 22	6.07	0.28	6.13	6.07	0.31	5.55	6.49
LingObf	8 22	-0.25	1.77	-0.30	-0.15	1.56	-5.84	2.97
cause	9 22	2.73	1.04	2.55	2.67	1.05	1.18	4.91
emo_pos	10 22	0.12	0.26	0.06	0.06	0.09	0.00	1.23
article	11 22	9.13	2.45	8.61	8.95	2.64	5.91	14.47
prep	12 22	15.66	1.15	15.89	15.83	0.76	12.82	17.22
quantity	13 22	4.66	1.64	4.46	4.55	1.43	2.05	8.66
${\tt PaperType}$	14 22	3.00	0.00	3.00	3.00	0.00	3.00	3.00
	range	skew k	ırtosis	se				
year	42.00	-1.11	-0.07	2.61				
Refs	134.00	1.32	1.43	6.94				
flesch_re	65.86	-1.16	0.90	3.35				
WC	8529.00	0.42	-0.93 !	516.97				
abstraction	6.82	-0.42	-0.84	0.40				
jargon	19.40	-0.30	-1.34	1.32				
CertSent	0.94	-0.27	-1.13	0.06				
LingObf	8.81	-1.03	2.29	0.38				
cause	3.73	0.42	-0.61	0.22				
emo_pos	1.23	3.38	11.22	0.06				
article	8.56	0.48	-0.91	0.52				
prep	4.40	-1.22	0.76	0.25				
${\tt quantity}$	6.61	0.62	-0.27	0.35				
PaperType	0.00	NaN	NaN	0.00				

-----

group: Multi-Authored Genuine Papers									
	vars	n	mean	sd	median	trimmed	mad	min	max
year	1	22	2010.77	8.50	2012.50	2011.89	8.15	1990.00	2022.00
Refs	2	22	44.05	20.05	36.50	41.94	15.57	19.00	91.00
flesch_re	3	22	34.86	8.34	36.17	35.12	6.35	15.95	49.45
WC	4	22	4407.82	2741.91	3624.00	4009.67	1307.65	1124.00	13901.00
${\tt abstraction}$	5	22	-0.56	1.83	-0.42	-0.48	1.65	-4.46	3.07
jargon	6	22	33.61	7.19	35.03	33.51	6.42	20.86	46.67
CertSent	7	22	6.19	0.31	6.22	6.20	0.19	5.21	6.76
LingObf	8	22	0.02	1.69	-0.45	-0.16	1.05	-2.50	4.28
cause	9	22	2.90	1.30	2.44	2.83	0.79	1.04	5.56
emo_pos	10	22	0.04	0.05	0.04	0.04	0.05	0.00	0.17
article	11	22	7.88	1.62	7.71	7.78	1.06	4.22	12.06
prep	12	22	15.48	2.15	15.66	15.58	1.41	9.76	20.67
quantity	13	22	3.75	1.09	3.74	3.76	1.21	1.69	5.53
PaperType	14	22	4.00	0.00	4.00	4.00	0.00	4.00	4.00
	ra	ange	e skew l	kurtosis	se				
year	32	2.00	-0.95	0.15	1.81				
Refs	72	2.00	0.77	-0.61	4.27				
flesch_re	33	3.50	-0.41	-0.37	1.78				
WC	1277	7.00	1.90	3.92	584.58				
${\tt abstraction}$	•	7.53	3 -0.26	-0.37	0.39				
jargon	2	5.81	-0.16	-1.01	1.53				
CertSent		1.55	-1.02	2.09	0.07				
LingObf	(	6.78	3 1.08	0.44	0.36				
cause	4	4.52	0.65	-0.77	0.28				
emo_pos	(	0.17	0.88	0.00	0.01				
article	-	7.84	0.52	0.94	0.35				
prep	10	0.91	-0.37	1.33	0.46				
quantity	;	3.84	-0.05	-1.19	0.23				
PaperType	(	0.00	) NaN	NaN	0.00				

### • Some quick notes:

- Year: The year seems to be relatively similar between groups, with the single-author groups (SAGP and SAFP) both having mid-2008 as the mean and the multi-author groups both having mid-2010 as the mean. This may be partially due to there being single outliers (matched papers that are both outliers) at the lower end of the distribution.
- Refs: As I suspected, the outlier of 146 seems to be pulling the mean for SAGP up to 51.05 (the means for all other groups are around 44). This is corroborated by the skew statistics which show that only the SAGP group has a skew value above

- 1. I may run the analyses where references is the outcome variable both with and without this case in the model, or I may use word count as a covariate in the model to try to account for it.
- WC: For some reason, there seems to be a notable difference in mean word count between the single-authored fraudulent group and the rest of the groups. I could speculate as to why this is the case, but I may do an unplanned follow-up analysis of this.
- 2. Frequencies will now be produced for categorical variables, both for the data in general and within PaperType groups.
- First, we will make the variables inst\_pres, gender, simple\_reason, Country, and PaperType factor variables.

```
# Changing inst_pres (institutional prestige), gender, simple_reason, Country, and PaperTy
data$inst_pres <- factor(data$inst_pres, levels = c(0, 1), labels = c("Not Major Research
data$gender <- factor(data$gender, levels = c("FEMALE", "MALE"), labels = c("Female", "Mal
data$simple_reason <- factor(data$simple_reason, levels = c("f_data", "f_image", "m_image"
data$Country <- factor(data$Country)
data$PaperType <- factor(data$PaperType)</pre>
```

• Next, we will produce frequency tables for each categorical variable of interest for the whole dataset.

```
# Creating the frequency tables for each categorical variable
freq_tab_inst_pres <- table(data$inst_pres)
freq_tab_gender <- table(data$gender)
freq_tab_simple_reason <- table(data$simple_reason)
freq_tab_Country <- table(data$Country)
freq_tab_PaperType <- table(data$PaperType)

# Displaying the frequency tables
freq_tab_inst_pres</pre>
```

Not Major Research Institution Major Research Institution

```
freq_tab_gender
Female Male
```

15 73

#### freq\_tab\_simple\_reason

Fabricated/Falsified Data Fabricated/Falsified Image 27 3

Manipulated Image Fabricated/Falsified Data and Image 22

### freq\_tab\_Country

Australia	Belgium	China	Egypt	Ethiopia
1	1	10	3	2
India	Iran	Israel	Italy	Japan
10	1	1	1	3
Latvia	Malaysia	Netherlands	Pakistan	Poland
1	2	6	3	2
Portugal	South Africa	South Korea	Taiwan	Turkey
2	1	2	1	2
United Kingdom	United States			
4	29			

#### freq\_tab\_PaperType

# MAFP MAGP SAFP SAGP 22 22 22

- Let's take a look at the way these are split:
  - First, there seems to be a preponderance of lower status (not major) research institutions in the whole dataset (70:18).
  - Next, most of the papers were retracted due to fabricated/falsified data (n = 27), followed by those that manipulated images (n = 12), fabricated/falsified images (n = 3), or both fabricated/falsified data and images (n = 2).
  - For country, the most common category of papers seem to be from the United States (n=29), followed by China and India (both n=10) and the Netherlands (n=6), with no other countries totaling more than n=4.
  - Finally, we can see that the papers are evenly split between each of our groups (n = 22; N = 88).

- Next, we will produce the frequencies within PaperType groups. Because we were not able to perfectly match across our matching characteristics, we will take this opportunity to look at how far off we were. A proportion table will be produced to more easily compare frequencies across PaperType groups.
  - First, we need to make the matching dummy variables into factor variables with labels to make them more interpretable.

```
# Creating factor variables out of the dummy variables that track matching across groups data$different_country <- factor(data$different_country, levels = c(0, 1), labels = c("Same data$different_gender <- factor(data$different_gender, levels = <math>c(0, 1), labels = c("Same data$different_inst_pres <- factor(data$different_inst_pres, levels = <math>c(0, 1), labels = c("Same data$different_journal <- factor(data$different_journal, levels = <math>c(0, 1), labels = c("Same data$different_journal]
```

• I have checked the dataset, and the labels seem to have been assigned correctly.

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