

# Stream of Consciousness Literature: A ‘Joyceless’ Linguistic Landscape\*

Exploring Word Frequency, Sentiment Value and Mental Health Themes in the  
Works of Joyce, Woolf, Proust, Mansfield and Eliot from Project Gutenberg

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This project focuses on understanding the language used by renowned stream of consciousness (SOC) authors James Joyce, Virginia Woolf, Marcel Proust, Katherine Mansfield and T.S Eliot. By conducting word frequency analysis and sentiment analysis on these authors’ nine novels, this paper aims to uncover shared linguistic patterns and gain insights into the authors’ mental states. This paper attempts to offer insights into themes of self-identity, anxiety, disassociation and existential contemplation within Western society and its literary circle from late 19th to mid-20th century. (add one sentence on main results)

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\*Code and data are available at: <https://github.com/ponolite/stream-consciousness-language.git>

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# 1 Introduction

Stream of consciousness (SOC) is a narrative technique that aims to capture the continuous flow of thoughts, feelings, and sensations experienced by a character without conventional organization or punctuation (Bernini and Fernyhough 2022). It mirrors the unpredictable and interconnected nature of human thought processes, often revealing the inner workings of the character’s mind in an intimate and unfiltered manner (Long and So 2016). In literature, most scholars agree that stream of consciousness reveals the complexities of mind-scapes, shedding light on the nuances of characters’ emotional well-being and psychological struggles (Nyongesa 2023). As such, this paper has mined the texts of a total of nine novels from the volunteer archive, Project Gutenberg, to examine the mental health themes of famous stream of consciousness authors, namely by Joyce, Woolf, Proust, Mansfield and Eliot, from the modernist

era of literature, spanning from late 19th century to mid-20th century (“Project Gutenberg,” n.d.). (more stats and data mentioned here)

By analyzing these textual datasets through word frequency and sentiment analysis, I seek to pose and answer crucial questions: *What are some important factors contributing to this relationship between mental health, disassociation and stream of consciousness? Moreover, how does this relationship vary differently across different demographics of authors, for instance, authors with different geographical locations and genders?* Understanding these dynamics is crucial in having an informed understanding of the West’s late 19th to mid-20th century literature and even socio-political landscape, especially in regards to how authors and creative writers navigate and deal with then-taboo topics such as existential angst, mental health issues and disabilities.

Thus, my estimand is the correlation between mental health-related words in SOC literature, their frequency and sentiment value as provided by Mohammad and Turney (2013). This is considered in terms of nine selected SOC novels only, namely Joyce’s *A Portrait of the Artist as a Young Man* and *Chamber Music*; Woolf’s *Mrs Dalloway* and *Jacob’s Room*; Proust’s *Swann Way*; Mansfield’s *Bliss* and *The Garden Party*; and Eliot’s *The Waste Land* and *The Love Song of J. Alfred Prufrock*. Through our analysis, we found that (percentage, number and data here, main results)...

To further understand the correlation between stream of consciousness novels and mental health themes, in [Introduction](#), the paper briefly discusses the nature of stream of consciousness literature, relevant authors and the works that I’ve chosen to analyze. Subsequently, in [Data](#) and [Results](#), I talk about the nature of the data obtained and analyze the results garnered from the data with suitable tables and charts. Next, [Discussion](#) provides further insights and future areas of study. Finally, [Conclusion] summarizes our main findings. To complete the paper, [Appendix](#) clarifies how each variable within each dataset is generated with relevant tables to accordingly demonstrate this.

The novel texts used for analysis were sourced from Project Gutenberg under the library `gutenbergr` (Johnston and Robinson 2023) (“Project Gutenberg,” n.d.). Data was generated, extracted and cleaned using the open-source statistical programming language R (R Core Team 2022), leveraging functions from `tidyverse` (Wickham et al. 2019), `tidytext` (Julia Silge and Robinson 2016), `rmarkdown` (Allaire et al. 2024), `dplyr` (Wickham et al. 2022), `ggplot2` (Wickham 2016), `scales` (Wickham, Pedersen, and Seidel 2023), `here` (Müller 2020), `igraph` (J. Silge and Robinson 2006), `widyr` (J. Silge and Robinson 2022), `ggraph` (Pedersen 2024), `textdata` (Hvitfeldt 2022), `tm` (Feinerer, Hornik, and Meyer 2008) and `knitr` (Xie 2014).

## 2 Data

### 2.1 Measurement

Two central variables in this paper are:

- **Word Frequency:** This variable captures the repetition of a single word (unigram), two-words combination (bigram) or three-words combination (trigram) throughout a SOC novel text, providing us with a thematic understanding of SOC literature.
- **Sentiment Value:** This variable enables us to analyze how every word is usually perceived emotionally, whether it be a qualitative feeling or if it is conveyed through a numerical value.

Out of two variables used, the first one, ‘Word Frequency’ usually captured as `n` or `frequency` in datasets, is directly quantified through tokenizing the novel texts using the package `tidytext` and its function `unnest_tokens()` (Julia Silge and Robinson 2016). To do this, I first downloaded all nine novel texts from “Project Gutenberg” (n.d.), and leveraged functions such as `unnest_tokens()` from Julia Silge and Robinson (2016) to mine the texts, or separating it into individual words. Finally, I used `count()` to quantify the word frequency.

The second variable used, ‘Sentiment Value’, is based on three English-based, general-purpose “word-emotion and word-polarity association lexicons”, sourced from Mohammad and Turney’s expansive research along with efforts from Finn Årup Nielsen and Bing Liu and collaborators (Julia Silge and Robinson 2016) (Mohammad and Turney 2013). The three general-purpose lexicon that contributes to my sentiment analysis are (Julia Silge and Robinson 2016):

- ‘AFINN’ from Finn Årup Nielsen, which assigns a numerical value from ‘-5 to 5’
- ‘bing’ from Bing Liu and collaborators, which assigns if a word is ‘positive’ or negative’
- ‘nrc’ from Saif Mohammad and Peter Turney, which assigns a core emotional value to a word, such as ‘fear’, ‘anger’, ‘sadness’ or ‘trust’.

In terms of measuring ‘Sentiment Value’, all three general-purpose lexicons are compiled through crowd-sourcing and directly surveying the public on how each word is emotionally perceived. A survey sample of how the word ‘startle’ is compiled within the ‘nrc’ lexicon is presented below (Mohammad and Turney 2013):

(1) Which word is closest in meaning (most related) to *startle*?

- automobile
- shake
- honesty
- entertain

(2) How positive (good, praising) is the word *startle*?:

- startle is not positive
- startle is weakly positive
- startle is moderately positive
- startle is strongly positive

(3) How negative (bad, criticizing) is the word startle?

- startle is not negative
- startle is weakly negative
- startle is moderately negative
- startle is strongly negative

After the survey results are garnered, researchers average the answers to sort each surveyed word into pre-defined categories, specifically ‘-5 to 5’ for ‘AFINN’, ‘postive’ or ‘negative’ for ‘bing’, and ‘anger’ or ‘fear’ for ‘nrc’. With the continuous work of compiling these lexicons spanning years and decades (Mohammad and Turney 2013) comes these functions: `get_sentiments("bing")` and `get_sentiments("nrc")` and `get_sentiments("afinn")` in which I can use `inner_join` to categorize my existing datasets of novel texts into their sentiment value. Systematic and data-driven, these measurement methods ensure that all lexicons faithfully reflects each word’s emotionality.

However, I do recognize how decontextualized and reductive this quantification of language can be. When it comes to understand such social and human artifacts as language or literary texts, much is dependent on their contextuality. As such, I will further discuss these weaknesses of the datasets under [Discussion](#).

## 2.2 Source Data: Data Cleaning and Word Tokenization

Discuss the authenticity/validity of Project Gutenberg

### 2.2.1 Word Count

Table 1: An Exemplary Table Containing Unprocessed Novel Text (James Joyce)

Book ID	Text	Book	Author
2817	To deep and deeper blue,	Chamber Music	James Joyce
2817	NA	Chamber Music	James Joyce
2817	III At that hour when all things have repose,	Chamber Music	James Joyce

Table 1: An Exemplary Table Containing Unprocessed Novel Text (James Joyce)

Book ID	Text	Book	Author
2817	O lonely watcher of the skies,	Chamber Music	James Joyce
2817	NA	Chamber Music	James Joyce

Table 2: An Exemplary Table Containing Tokenzied Novel Text (James Joyce)

Book ID	Book	Author	Word
4217	A Portrait of the Artist as a Young Man	James Joyce	stead
4217	A Portrait of the Artist as a Young Man	James Joyce	dublin
4217	A Portrait of the Artist as a Young Man	James Joyce	1904
4217	A Portrait of the Artist as a Young Man	James Joyce	trieste
4217	A Portrait of the Artist as a Young Man	James Joyce	1914

Table 3: An Exemplary Table Containing Word Count of Each Word within Novel Texts (James Joyce)

Word	Count
stephen	373
god	194
eyes	180
soul	178
father	151

### 2.2.2 Comparative Word Frequency

Table 4: Word Frequency of Stream of Consciousness Novels, A Comparison Between Five Authors

Word	James Joyce	Katherine Mansfield	Marcel Proust	T.S. Eliot	Virignia Woolf
abandon	0.000112	NA	8.19e-05	NA	NA
abandoned	0.000112	NA	8.19e-05	NA	NA
abandonment	0.000112	NA	8.19e-05	NA	0.0001421
abase	0.000112	NA	NA	NA	NA
abased	0.000112	NA	NA	NA	NA

### 2.2.3 Generating Word Networks

## 2.3 Data Limitations

## 3 Model

The goal of our modelling strategy is twofold. Firstly,...

Here we briefly describe the Bayesian analysis model used to investigate... Background details and diagnostics are included in Appendix [6.2](#).

### 3.1 Model set-up

Define  $y_i$  as the number of seconds that the plane remained aloft. Then  $\beta_i$  is the wing width and  $\gamma_i$  is the wing length, both measured in millimeters.

$$y_i | \mu_i, \sigma \sim \text{Normal}(\mu_i, \sigma) \quad (1)$$

$$\mu_i = \alpha + \beta_i + \gamma_i \quad (2)$$

$$\alpha \sim \text{Normal}(0, 2.5) \quad (3)$$

$$\beta \sim \text{Normal}(0, 2.5) \quad (4)$$

$$\gamma \sim \text{Normal}(0, 2.5) \quad (5)$$

$$\sigma \sim \text{Exponential}(1) \quad (6)$$

We run the model in R (R Core Team 2023) using the `rstanarm` package of (`rstanarm?`). We use the default priors from `rstanarm`.

### 3.1.1 Model justification

We expect a negative relationship between average household income and the number of children per child care space by ward. In particular...

We can use maths by including latex between dollar signs, for instance  $\theta$ .

## 4 Results

### 4.1 The Dominant Vocabulary of SOC Literature



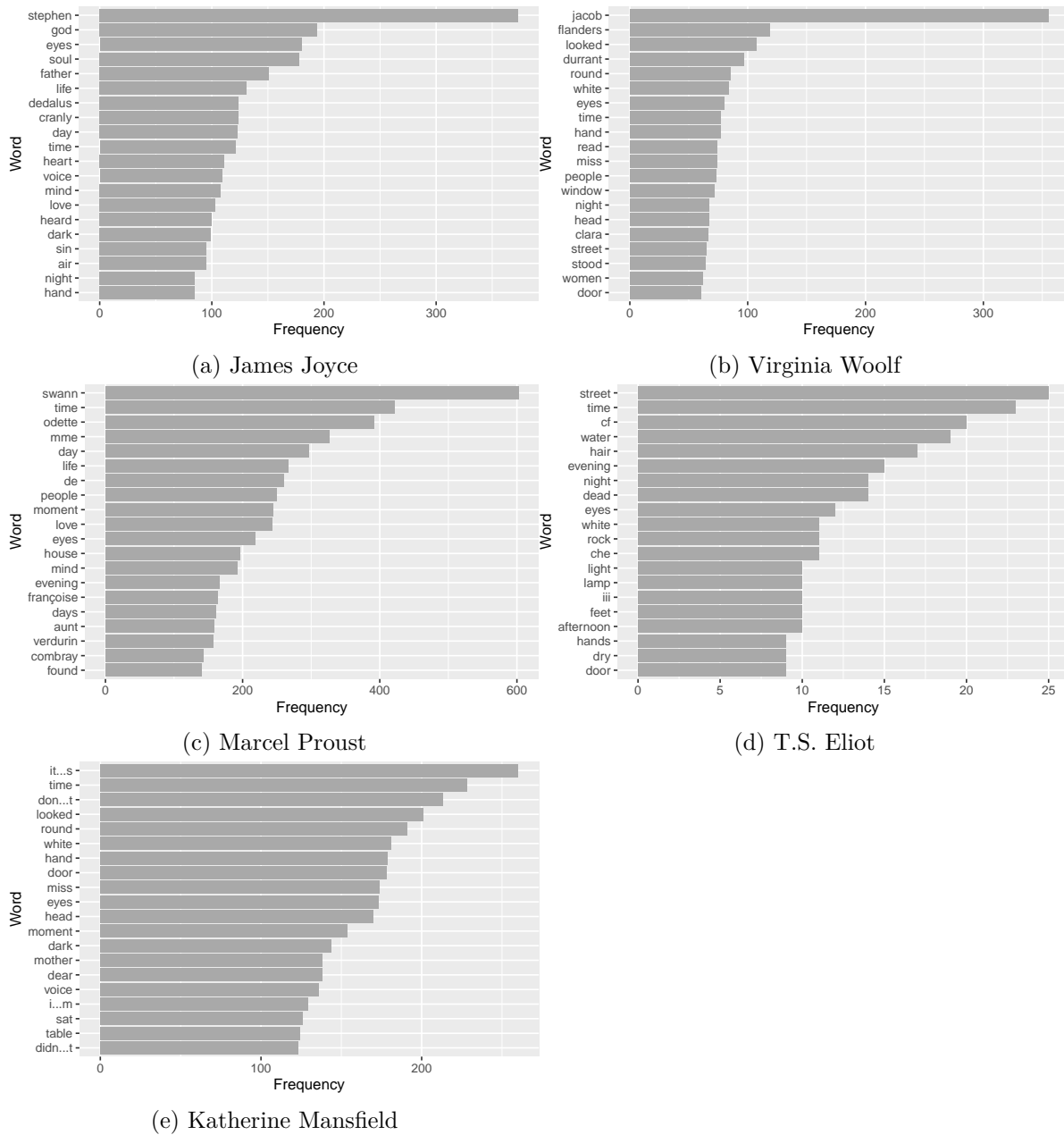


Figure 1: Comparative Analysis of Top 20 Word Frequencies from Famous SOC Authors

## 4.2 Sentiment Analysis

Leveraging sentiment analysis from Mohammad and Turney (2013)

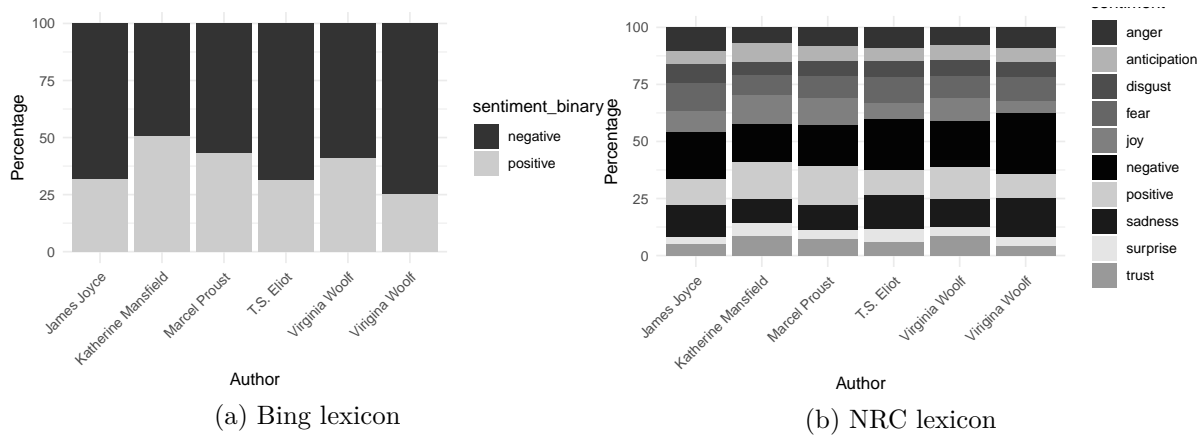


Figure 2: Categorical Sentiment Analysis of All SOC Novel Texts by Authors

## 4.3 Gendered Mental Landscape of Stream of Consciousness Novels

### 4.3.1 Comparing SOC Literature's Female Authors

### 4.3.2 Comparing SOC Literature's Male Authors

## 4.4 Transnational SOC Novels and Mental Health Themes

## 4.5 Combined Texts: Trends, Word Networks, Bigram and Trigram Analysis

# 5 Discussion

## 5.1 Mental Health Vocabulary: Patterns and Trends

Discuss vocabulary patterns and word trends

## 5.2 Insights into Socio-Political Landscape of the West's Modernist Era

The novels' linguistic patterns reflect the socio-political landscape of the Western hemisphere, from late 19th century to the mid-20th century.

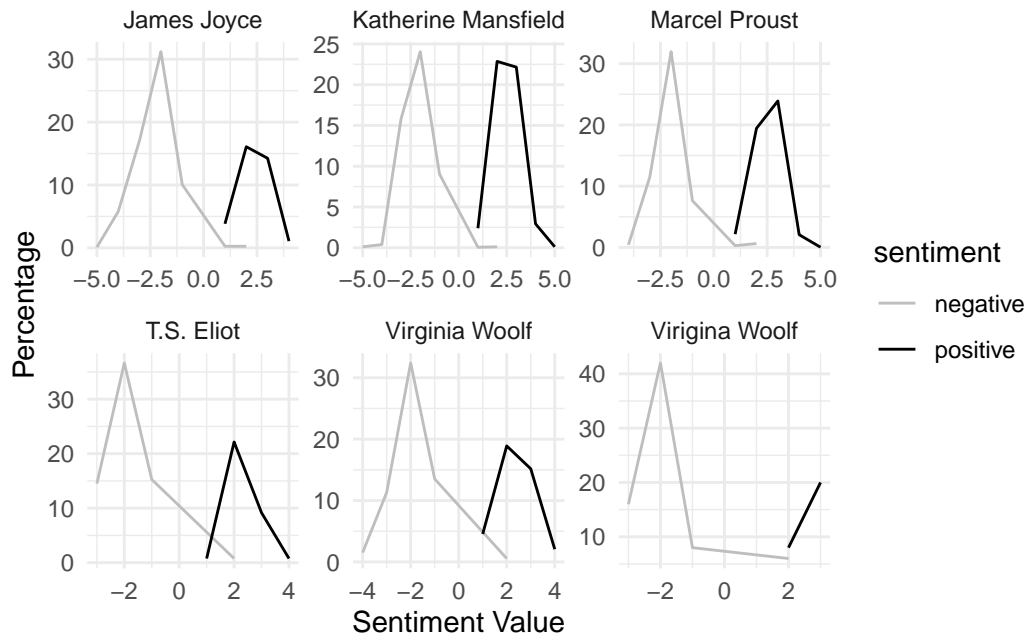


Figure 3: Numerical Sentiment Analysis of All SOC Novel Texts by Authors (AFINN lexicon)

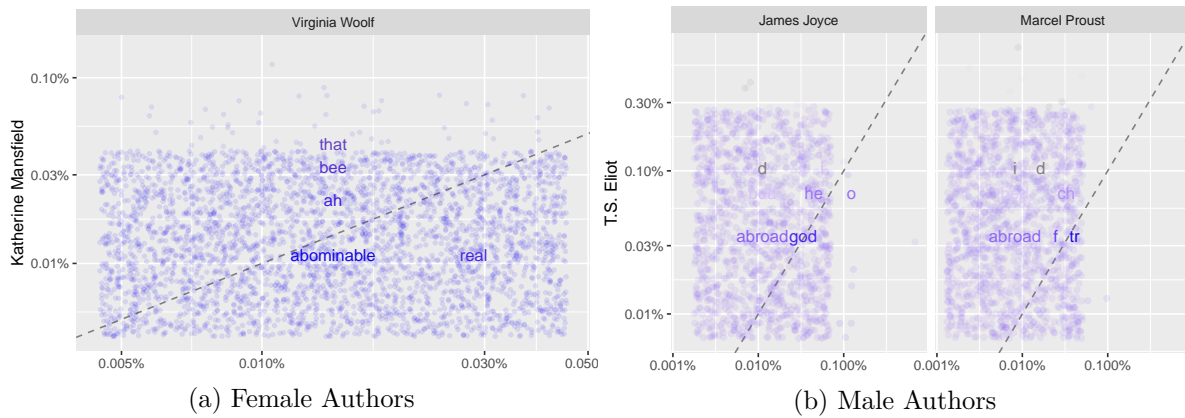


Figure 4: Comparative Analysis of Word Frequency in Female and Male Stream of Consciousness Authors

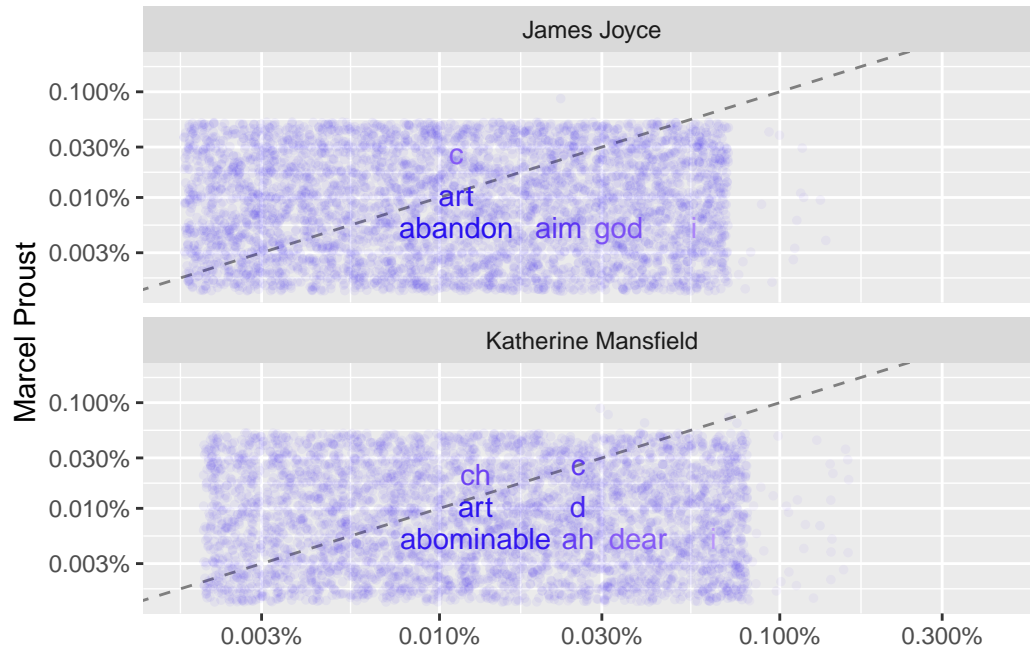


Figure 5: Comparative Analysis of Word Frequency in Transnational Stream of Consciousness Authors

### 5.3 Schizophrenic and Disassociative Tendencies in Female Stream of Consciousness

#### 5.4 Weaknesses

##### 5.4.1 Lack of Thorough Word Cleaning

This includes words such as chapter titles, Roman numbers and personal names, affecting the integrity of data analysis (“Swann” being the most common word)

##### 5.4.2 Decontextualized Literature Works and Limiting Publication Editions

Words are singled out and analyzed without context which could have affected their intended meanings, especially in such a complex genre such as stream of consciousness. The limiting novel editions also doesn’t make sure that their literary integrity are maintained and the analysis might have missed important texts of other editions.

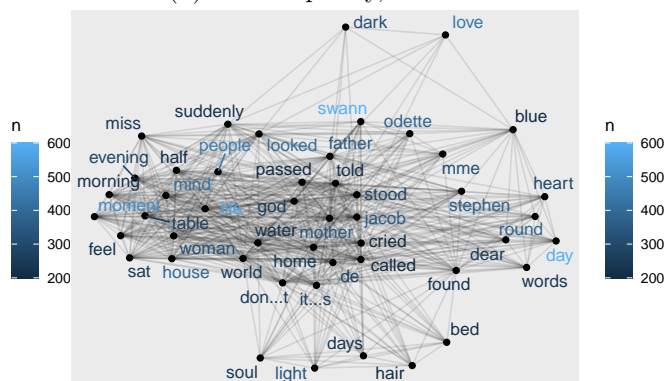
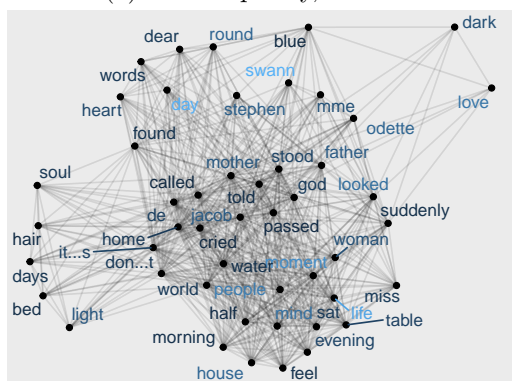
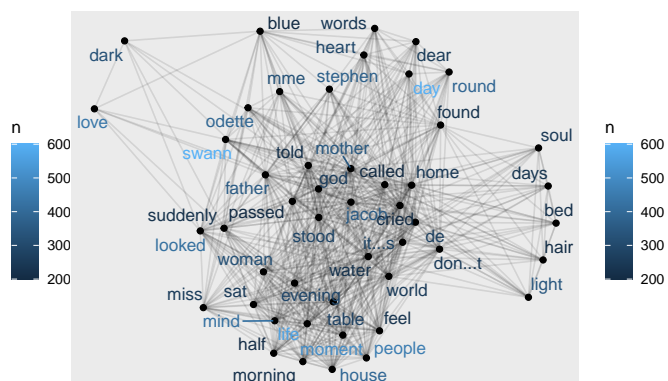
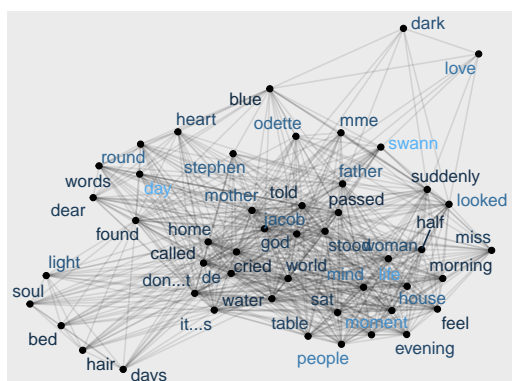


Figure 6: Word Networks Measured by Frequency and Correlation when Combining All Stream of Consciousness Novels

### **5.4.3 Uneven Novel Length and Categorization of Authors**

Since the novels are chosen based on their worldwide reception and canon, practical constraints like how all novels should be of similar length are ignored. This constraint can prevent one novel having more words than others, which can critically affect the integrity of the word frequency analysis, sentiment value and word networks where one novel dominates the others and skews the results.

In addition, these authors work are expansive and so choosing a select few to bind them to the SOC genre can be limiting as the SOC genre in itself is already an amalgamation of different literary trends. Thus, this can affect the integrity of the datasets.

### **5.4.4 Project Gutenberg's Focus on the Canon**

Along the search for other stream of consciousness authors to include in the generated datasets, only a few-those that are in the SOC canon, mostly white authors-are present in Project Gutenberg, without including lesser known SOC authors, which could have steared the data analysis, still, towards their preconception: being rueful and angst-ridden

## **5.5 Moving Forward and Next Steps**

## 6 Appendix

### 6.1 Additional Data Details

```
##| eval: true  
##| echo: false  
##| message: false  
##| warning: false  
#combined_books
```

#### 6.1.1 Data Gathering

```
##| echo: false  
##| message: false  
##| label: tbl-reasons-strip-search  
##| tbl-cap:
```

#### 6.1.2 Data Cleaning

```
##| echo: false  
##| message: false  
##| label: tbl-items-strip-search  
##| tbl-cap:
```

### 6.2 Model Details

### 6.3 Posterior predictive check

### 6.4 Diagnostics

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