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Measuring and Mapping Intergeneric Allusion in Latin Poetry using Tesseræ

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Abstract

Most intertextuality in classical poetry is unmarked, that is, it lacks objective signposts to make readers aware of the presence of references to existing texts. Intergeneric relationships can pose a particular problem as scholarship has long privileged intertextual relationships between works of the same genre. This paper treats the influence of Latin love elegy on Lucan's epic poem, *Bellum Civile*, by looking at two features of unmarked intertextuality: frequency and distribution. I use the Tesseræ project to generate a dataset of potential intertexts between Lucan's epic and the elegies of Tibullus, Propertius, and Ovid, which are then aggregated and mapped in Lucan's text. This study draws two conclusions: 1. measurement of intertextual frequency show that the elegists contribute fewer intertexts than, for example, another epic poem (Virgil's *Aeneid*), though far more than the scholarly record on elegiac influence in Lucan would suggest; and 2. mapping the distribution of intertexts confirms previous scholarship on the influence of elegy on the *Bellum Civile* by showing concentrations of matches, for example, in Pompey and Cornelia's meeting before Pharsalus (5.722-815) or during the affair between Caesar and Cleopatra (10.53-106). By looking at both frequency and proportion, we can demonstrate systematically the generic enrichment of Lucan's *Bellum Civile* with respect to Latin love elegy.

Keywords

allusion; intertextuality; generic enrichment; Lucan; Latin epic; Latin love elegy; Tesseræ

I INTRODUCTION

“There is no better way to penetrate the secrets of Lucan's workshop, to observe how the poem crystallized in his mind, than to examine passages where he borrows from, adopts, or echoes his predecessors.” So wrote [Bruère, 1951 p. 222]. Bruère's contributions to the study of Lucan's allusive practice, especially his two articles co-authored with Thompson ([Thompson and Bruère, 1968; Thompson and Bruère, 1970]), brought a heightened awareness of the intertextual nature of the *Bellum Civile*, and in particular, its relationship to Virgil's *Aeneid*. The presence of intertextual influence from genres other than epic, however, has received far less attention. Recent work by [Sannicandro, 2010; Caston, 2011; McCune 2014] has sought to remedy this imbalance. These studies, however, have tended to emphasize localized readings, either referring to a limited number of elegiac source texts or treating a select group of episodes in Lucan's poem. In this paper, I use datasets drawn from the Tesseræ Project at the University at Buffalo ([http1]) to systematically compare potential intertexts between Lucan's epic as a whole with reference to the complete works of the Latin love elegists Tibullus, Propertius, and Ovid.

II BACKGROUND

Systematic collections of elegiac references, no less analysis of these references, in epic poetry remain a desideratum in Latin literary criticism. Even within the epic genre, there are two works of traditional philological research which stand out for treating influence in a systematic and comprehensive manner, namely [Knauer, 1964] (with its subtitle “mit Listen

der Homerzitate in der Aeneis”) and [Nelis, 2001]. [Farrell, 2005 p. 107] has written that the “mind recoils from the thought of a library full of books entitled, ‘The *Aeneid* and Homer,’ ‘The *Aeneid* and Apollonius,’ ‘The *Aeneid* and Ennius,’ and so forth.” At the same time, having access to this kind of reference material would undoubtedly be useful. The existence of a book called “The *Bellum Civile* and Latin Love Elegy” would certainly appear in the bibliography of this study if it existed. [Coffee et al, 2012] remarks that traditional scholarly methods have avoided these kinds of comprehensive treatments of intertextuality because of the massive scholarly labor involved. Software is now available, however, to greatly reduce the procedural difficulty to which Coffee refers. Lists of potential intertexts can be compiled much more easily using Tesseract’s web based tool ([http1]), which allows for the quick gathering of evidence for potential intertextuality between two texts, shifting scholarly labor from detection to analysis.

2.1 Problem of unmarked intertextuality

In his study of the intertextual relationship between Horace and Lucan, [Groß, 2013] observes that almost all intertextuality in classical poetry is unmarked, that is, it is not characterized by explicit signposts, but rather through implicit markers. Here, Groß follows the definition of unmarked intertextuality from [Helbig, 1996], who includes as two of the implicit markers 1. the “frequency” (*Frequenz*) of intertexts in the target text, and 2. the “distribution” (*Distribution*) of intertexts throughout the text. This definition finds sympathy in two literary critical approaches to the problem of unmarked intertextuality, namely the “allusive system” discussed by [Farrell, 2005] and the “code model” discussed by [Conte, 1986]. Both Farrell and Conte argue that the relationship between two texts can be drawn to some degree by the volume of potential intertexts and their consistent presence throughout a target text. A collocation tool like Tesseract, by algorithmically determining and reporting a complete collection of correspondences, offers a formalization of Farrell’s system and Conte’s model. Moreover, the data collected from Tesseract results can be used to formalize Helbig’s observation about frequency and distribution as implicit signposts for unmarked intertextuality. The analysis of Tesseract results can measure frequency by showing the number of times similarity in word use triggers a match and can measure distribution by showing which parts of the *Bellum Civile* show a greater or lesser number of matches.

2.2 Literature Review

In recent years, researchers at Tesseract have published a series of papers testing the assumptions of traditional Latin literary criticism against their algorithmic model ([Coffee et al, 2012; Coffee et al, 2013; Forstall et al, 2015]). These papers have used the first book of Lucan’s *Bellum Civile* as their target text and Virgil’s *Aeneid* as their source text, evaluating the results of the automated tool against philological commentaries by assigning them, following [Thomas 1986], values of “meaningful” and “not meaningful,” as well as “interpretable” and “not interpretable.” [Forstall et al, 2015] reports that scores assigned by the Tesseract algorithm correlate well with supervised assignments of meaning and interpretability.

The Tesseract publications have confirmed the traditional scholarly view that Lucan’s poetic diction draws significantly on Virgil. That said, this research has consistently pointed the way towards wider applicability of algorithmically based methods for the study of intertextuality: [Coffee 2012] suggests that systematic collection and measurement of textual similarities using a tool like Tesseract can build an “intertextual ‘fingerprint,’” that can be used to make meaningful comparisons between the poetic practices of different authors.

Important work on testing Tesseract search results is also being done by [Bernstein, 2013; Gervais, 2014; Bernstein and Gervais, 2015], who have concentrated on the platform’s “macrophilological applications,” that is ways in which the complete collection of search results for a given genre, author, or work can be used to draw conclusions, not about specific intertexts, but rather about larger patterns of intertextuality. [Bernstein and Gervais, 2015], in particular, in a study that looks at intertextual relationships in Latin hexameter poetry as a whole, argues that Tesseract can be used to generate an unsupervised dataset which captures the intertextual relationship between multiple Latin texts and can then be used as the basis for further analysis and interpretation.

III DATA

3.1 Texts

This study uses the following texts available from the Tesseract Github repository ([http2]).

The following editions of Latin epic poetry are used:

- Virgil, *Aeneid*: Greenough, J. B., ed. 1900. *Bucolics, Aeneid, and Georgics of Vergil*. Boston: Ginn & Co.
- Lucan, *Bellum Civile*: Weise, H., ed. 1835. *M. Annaei Lucani Pharsaliae Libri X*. Leipzig: G. Bassus.

The following editions of Latin love elegy are used:¹

- Tibullus: Postgate, J.P., ed. 1915. *Tibulli aliorumque Carminum libri Tres*. Oxford: Oxford University Press.
- Propertius: Mueller, L., ed. 1898. *Sex. Propertii Elegiae*. Leipzig: Teubner.
- Ovid: Ehwald, R., ed. 1907. *Amores, Epistulae, Medicamina faciei femineae, Ars amatoria, Remedia amoris*. Leipzig: Teubner.

I have used the volumes listed above so that meaningful comparisons can be made with Tesseract studies which have already been published as well as those being conducted by other researchers. This follows the recommendation of [McGillivray, 2014], who argues that it is methodologically critical to work within a “collaborative research paradigm,” that is to work from a common set of texts and to build directly upon existing tools and frameworks in an effort to maintain replicability in literary research. I have published the data set and the code used to generate the tables and figures on Github [http3].

3.2 Tesseract Search Results

¹The collection listed above has been decided upon in order to align this work with that of Tesseract. It is obviously not the only arrangement available. [Pichon, 1902], for example, defined his sample as follows: the canonical works of Latin elegy mentioned in Ovid *Tristia* 4.10.53-54 and Quintilian *Institutiones* 10.1.93, to which he adds (or qualifies the inclusion of) Catullus, the *Corpus Tibullianum*, all of the *Heroides* regardless of authenticity, and certain poems from Ovid’s *Tristia* and *Epistulae ex Ponto*. For Ovid, I use Ehwald’s editorial decision to define the subset of Ovid’s elegiac work which qualifies as erotic. Accordingly, the *Fasti*, *Ibis*, *Tristia* and *Epistulae ex Ponto* will not be used in this study. Along similar lines, I have based my decision to include the *Corpus Tibullianum* on Postgate’s editorial decision and Tesseract’s use of this edition.

Tesseractae describes itself as a framework for “detecting allusions” in Latin poetry. More precisely, it is a search tool designed: 1. to compare the texts of two authors by looking for shared words, and 2. to return a list of similar passages scored for significance on a scale from 2 to 10. All results require that the units of text under consideration contain a minimum of two shared words. Once this requirement is met, matches are scored algorithmically based on two factors: word frequency and phrase density. Descriptions of the scoring algorithm can be found in [Forstall et al, 2015]. Word frequency refers to how common or uncommon a matched word is within the corpus. Phrase density refers to the distance between matched words in the two texts. These parameters are designed to make explicit the formal criteria that scholars have traditionally applied implicitly when identifying an allusion. Accordingly, less common words which are adjacent receive higher scores than more common words which are separated by gap of several words. For example, the adjacent collocation of the rare words *livor edax* (Ovid *Amores* 1.15.1: *Quid mihi Livor edax, ignavos obicis annos... ~ Lucan Bellum Civile* 1.288: *Livor edax tibi cuncta negat: gentesque subactas*) receives a score of 10, while a separated collocation of the very common words *quod* and *te* (e.g., Ov. *Amores* 2.9b.47 ~ Lucan *Bellum Civile* 9.854) receives a score of 3.

On the one hand, by its nature, the Tesseractae algorithm collects matches in an unsupervised manner. That is, it returns as many matches as fit its criteria—the text analytic equivalent of trawl fishing or strip mining. The result is a high number of false positives or dubious, semantically empty connections, especially due to ambiguity in lemmatization ([Forstall et al, 2015]). These matches correlating with the low end of the Tesseractae scoring scale will be largely ignored in this study. On the other hand, [Coffee et al, 2012; Forstall et al, 2015] have shown that the high scores (that is, scores 10 and 9) generated by the Tesseractae scoring algorithm correlate with meaningful and interpretable results, and that the next tier of scores (scores 8 and 7) correlate with meaningful results. Accordingly, the high scores will be the focus of this study.

The data used for the study of intertextual frequency and distribution was gathered using version 3 of the Tesseractae search interface. The following parameters were used for these searches:

- unit = line
- feature = stem
- stopsize = 20
- stbasis = corpus
- stopwords = qui quis sum et in is hic non ego ut cum tu ad ille quod ab si atque neque sed
- max_dist = 10
- dibasis = freq
- cutoff = 0

Full definitions of the terms used in the list below are available at the Tesseractae site [http4].² Defaults were used where possible to ensure to the greatest degree possible

²The following comments should be helpful in getting a quick understanding of the Tesseractae settings. “Feature” determines how words are treated by the algorithm for making comparisons; under the setting “stem,” lemmas are matched, that is, *amor* in one text matches *amoris*, *amori*, *amore*, etc. in another. “Stopwords” is the default Tesseractae stoplist, that is, the list of words ignored in this study, based on the 20 most commonly appearing words in the corpus of Latin literature used by the platform. “Max_dist” refers to the maximum distance that the algorithm uses for its window for matches; that is words in a text must be between two and ten words apart from

comparability between this study and other Tesserae-based studies. One exception is “line” as the unit for this study; the maps of intertextual distribution use line numbers for the x-axis and the number of matches per line for the y-axis.

Tesserae results yield the following information:

- Result number
- Target Location
- Target Text
- Source Location
- Source Text
- Shared words
- Score

Here is an example of a record from the .csv file returned by a Tesserae search between the *Bellum Civile* and Ovid’s *Amores*: 11, “luc. 1.288”, “**Livor** **edax** tibi cuncta negat: gentesque subactas”, “ov. am. 1.15.1”, “Quid mihi **Livor** **edax**, ignavos obicis annos”, “edax; liuor”, 10

IV METHODS

Following [Helbig, 1996], I have designed this study to measure two of his criteria for unmarked intertextuality: frequency and distribution.

Frequency is measured by aggregating the count of matches by author or work. Tesserae returns search results as a .csv and these results are then converted into dataframes for processing with the Python Pandas module ([McKinney, 2012]). Raw counts are normalized to matches per 100 lines to account for differing text lengths. Lengths were determined using the Tesserae texts given in the section “2.1 Texts” above. Tesserae scores below 7 are discarded to reduce “noise” and to restrict the analysis to scores which have been shown to yield the most significant results ([Forstall et al, 2015; Bernstein and Gervais, 2015]).

Distribution is measured by using Pandas to aggregate the total number of Tesserae matches from the elegists above score 7 for each line in the *Bellum Civile*. Lines from the *Bellum Civile* with no matches from Tesserae are assigned a count of zero. Because the goal of studying the distribution of unmarked intertextuality is to see how the feature appears throughout the entire work, rather than in any given line, the counts of matches per line are smoothed by taking the running average of scores within a specific window, here 25 lines. This reduces the line-by-line variability in counts while providing a better sense of potential intertextual density in different sections of the target text. The smoothed scores are then mapped by book using line numbers for the x-axis and the smoothed counts of Tesserae matches for the y-axis.

V DISCUSSION

5.1 Measuring Frequency

each other, counting inclusively, to yield a result. “Dibasis” refers to the manner in which the algorithm accounts for distance between words; according to [http4], with the setting “freq,” Tesserae “attempts to zero in on the most relevant words in an allusion, measuring the distance only between the phrase’s two most infrequent words.” “Cutoff” refers to the minimum score returned by Tesserae; I have set it at zero to gather the full range of Tesserae results, although scores below 7 will be dropped for most parts of this study as noted below.

Table 1 shows the total number of Tesseract matches with Lucan's *Bellum Civile* as the target text for each of the elegists and their works, with Virgil's *Aeneid* included for comparison. The raw counts, however, are not sufficient to compare the authors/works, because they are of greatly varying lengths. For example, Propertius's four books of elegies have a total of 3,982 lines compared to the 9,896 lines of Virgil's *Aeneid*. Accordingly, it is necessary to normalize these scores so that they can be compared more usefully. By normalizing the counts of Tesseract results to matches per 100 lines, we get a different picture than the received view of intertextuality in Lucan, as shown in Table 2. The number of matches between Lucan and Virgil is much higher than the number between Lucan and any of the elegists. When normalized, however, the elegists show on average only a 17% difference, and in the case of Ovid's *Heroides*, the matches per 100 lines is slightly higher (469.67 in the *Heroides* versus 468.72 in the *Aeneid*).

Figure 1 shows the total number of matches by score for Virgil. As noted above, research on Tesseract results shows that results scoring 7 or above correlate best with meaningful results. Accordingly, in order to reduce the signal-to-noise ratio and limit the remainder of this part of the study to only the most meaningful results, scores below 7 will not be considered below. Figure 2 shows the total number of matches by score above this threshold for Virgil.

By combining the strategies listed above of normalizing the scores and concentrating on the scores above a certain threshold, we can now compare the elegists to each other, again with the *Aeneid* included as a baseline. Looking at Figure 3, we can see that the shared diction between Virgil and Lucan as measured by the Tesseract algorithm, that is with consideration of word frequency and phrase density, is higher than for any of the elegists. This is hardly surprising considering that the history of scholarship on Lucan has privileged this relationship far beyond that of any other poetic predecessor. Rather the conclusion that can be drawn from this chart is that the minimal scholarly attention that has been paid to the influence of the elegists on Lucan, based on the Tesseract data, is out of proportion with the attention paid to the *Aeneid*.

The score most likely to be meaningful and interpretable, that is score 10, is exceptionally rare in all authors, consistently appearing less than one time per 100 lines. For score 9, the elegists show roughly half as many matches per 100 lines (average 2.77 per 100 lines; 48.7% of Virgil's 5.70). The count for Ovid's *Heroides* is notable for standing out as being much higher than the others (4.25 matches per 100 lines; 90.4% of Virgil's 5.70), suggesting that the density of specific allusions noted between these poems and the *Bellum Civile* by previous scholars (e.g. [Bruère 1951; Sannicandro 2010]) is supported by the evidence.

With the meaningful and non-interpretable scores (8 and 7), that is those which contribute to the general elegiac texture of the work and show evidence of elegy as a code model for the *Bellum Civile*, we see a similar pattern. Virgil's *Aeneid* again shows a higher number of matches per 100 lines, but not by the overwhelming margin that scholarly attention between the two genres would suggest: for score 8, the elegists show a little more than two-thirds as many matches per 100 lines (average 14.95 per 100 lines; 70.0% of Virgil's 21.38), and for score 7, more than three-quarters as many matches (average 68.0 per 100 lines; 77.5% of Virgil's 87.76).

5.2 Mapping Distribution

In addition to data about frequency, Tesseract searches for the elegists and the *Bellum Civile* also provide us with data about the location of the matches. Figure 4 shows a plot of the

distribution of aggregated Tesseract matches per line for the complete set of elegiac texts as the source text and book 1 of the *Bellum Civile* as the target text. We learn from this text map that the matches are randomly distributed throughout the book, suggesting that elegy is functioning as a code model for the *Bellum Civile*. The line with the highest number of matches (15) is Lucan 1.61 (...*inque vicem gens omnis amet: Pax missa...*), likely reflecting the presence of “love” (*amet*) in this line, the central theme of the elegiac poems under consideration; the average number of matches per line is 1.69, and 65% (449 out of 695) of lines in book 1 have at least one match.

That said, the elegiac weight of any given line gives us information that is perhaps too localized. In order to get a better sense of which episodes show a sustained interaction with elegy, it is preferable to plot the running average of matches within a certain window of lines. Figure 5 shows a plot of this running average for each book of the *Bellum Civile* using a window of 25 lines.

As compared to the line-by-line mapping of counts from the Tesseract results, plotting by window confirms once again that matches are distributed consistently throughout the poem. It also, and much more usefully as a prompt for further investigation, provides a clearer idea of where in the *Bellum Civile* additional research into elegiac influence would be most fruitful. Certain peaks in these text maps corroborate the work of previous scholarship on elegiac influence in Lucan. For example, we see pronounced upticks in the average number of matches at the end of book 5 for Pompey and Cornelia’s meeting before Pharsalus (Lucan 5.722-815; see [Sannicandro, 2010; Bruère, 1951]), and at the beginning of book 10 for the appearance of Cleopatra (Lucan 10.53-106, 172-192; see [McCune, 2014; Groß, 2013]). Another interesting takeaway from this view of the distribution of elegiac language in the *Bellum Civile*, is the presence of peaks in parts of the epic where scholarly research into the interaction of these two genres has not been previously focused, as, for example, in the book 1 proem (1.1-32) or Cato’s speech to his troops in book 9 (9.222-283).

VI Conclusion

The automated detection and measurement of intertextuality, in particular the definition of “allusion” formalized by Tesseract, offers the insight into the “poet’s workshop” that Bruère described. It allows researchers to systematize and quantify the intertextual readings brought out in traditional, qualitative literary analysis of Latin poetry. This study uses the evidence generated by Tesseract to support Helbig’s conception of frequency and distribution as implicit indicators of unmarked intertextuality. It also provides data to the idea that an “allusive system” can be deduced from a mass of textual similarities. The large number of scores 7 and higher for all three Latin love elegists, and the fact that these results are not confined to a small number of locations in the text suggest that this genre acts to some degree as a “code model” for the *Bellum Civile*.

[Coffee et al, 2013 p. 227] showed how Tesseract could be used to corroborate the dominant scholarly opinion about Lucan’s reliance on Virgil “for the basic idiom of epic.” What is less in line with the traditional view of Lucan’s allusive practice however is to what extent the results given above demonstrates elegy’s contribution to his epic “idiom.” The frequency of intertextual correspondence may be less than for Virgil’s *Aeneid* and their distribution more diffuse, but this study was not meant to disprove Virgil’s authority or deny his influence on the later epic poet. Rather it is meant to show that another genre can compete for some space in Lucan’s attention to poetic predecessors. Scholarly consideration of the effect of elegy on the *Bellum Civile* is miniscule compared to the massive amount of research on Lucan’s poetic

relationship to, or even dependence, on Virgil as a poetic model. Yet, as we see here, the gap in the evidence is not as wide as the record of scholarship suggests.

This distant reading approach to verifying the allusive system and the code model is a starting point. [Bernstein and Gervais 2015] comments that a limitation of the Tesseract platform is that it does not yet offer the “sensitive assessment of significance” that is taken for granted in traditional forms of Latin literary criticism. Along these lines, the test for frequency in this study does not address whether individual results are in fact meaningful or interpretable, and the test for distribution, while pointing out episodes worthy of further scrutiny, has not evaluated them. [Forstall et al, 2015] reminds users that Tesseract data must still be interpreted within the established scholarly conversation concerning intertextuality in Latin poetry. Examination of the peaks and troughs of the text maps of allusive distribution forms an excellent starting point for a more “sensitive assessment” of the intertextual relationship between Lucan and the love elegists in the future and also provides a more empirical method for investigating the “generic enrichment” ([Harrison, 2007; Harrison, 2013]) found in Latin poetry.

VII TABLES AND FIGURES

7.1 Tables

	Work	Tesseract Matches
Tibullus	<i>Elegies</i>	6,860
Propertius	<i>Elegies</i>	13,812
Ovid	<i>Amores</i>	9,966
	<i>Ars Amatoria</i>	9,235
	<i>Heroides</i>	18,646
	<i>Medicamina</i>	285
	<i>Remedia Amoris</i>	3,711
Virgil	<i>Aeneid</i>	46,385

Table 1. Total number of Tesseract matches for Lucan’s *Bellum Civile* as the target text and the Latin love elegists and Virgil’s *Aeneid* as source texts.

	Work	Lines	Tesseract Matches	Count per 100 Lines
Tibullus	<i>Elegies</i>	1,929	6,860	355.62
Propertius	<i>Elegies</i>	3,982	13,812	346.86
Ovid	<i>Amores</i>	2,445	9,966	407.61
	<i>Ars Amatoria</i>	2,330	9,235	396.35
	<i>Heroides</i>	3,970	18,646	469.67
	<i>Medicamina</i>	100	285	285.00
	<i>Remedia Amoris</i>	814	3,711	455.90
Virgil	<i>Aeneid</i>	9,896	46,385	468.72

Table 2. Number of Tesseract matches normalized per 100 lines for Lucan’s *Bellum Civile* as the target text and the Latin love elegists and Virgil’s *Aeneid* as source texts.

	Work	Score 10 (per 100)	Score 9 (per 100)	Score 8 (per 100)	Score 7 (per 100)
Tibullus	<i>Elegies</i>	0.31	2.64	18.56	76.31
Propertius	<i>Elegies</i>	0.33	3.16	15.77	67.88
Ovid	<i>Amores</i>	0.49	2.74	18.45	75.83

	<i>Ars Amatoria</i>	0.13	2.79	14.98	72.27
	<i>Heroides</i>	0.13	4.26	18.49	79.07
	<i>Medicamina</i>	1.00	2.00	6.00	33.00
	<i>Remedia Amoris</i>	0.25	1.84	12.41	71.62
Virgil	<i>Aeneid</i>	0.87	5.70	21.38	87.76

Table 3. Number of Tesseræ matches by score (at or above a threshold of 7) normalized per 100 lines for Lucan’s *Bellum Civile* as the target text and the Latin love elegists and Virgil’s *Aeneid* as source texts.

7.2 Figures

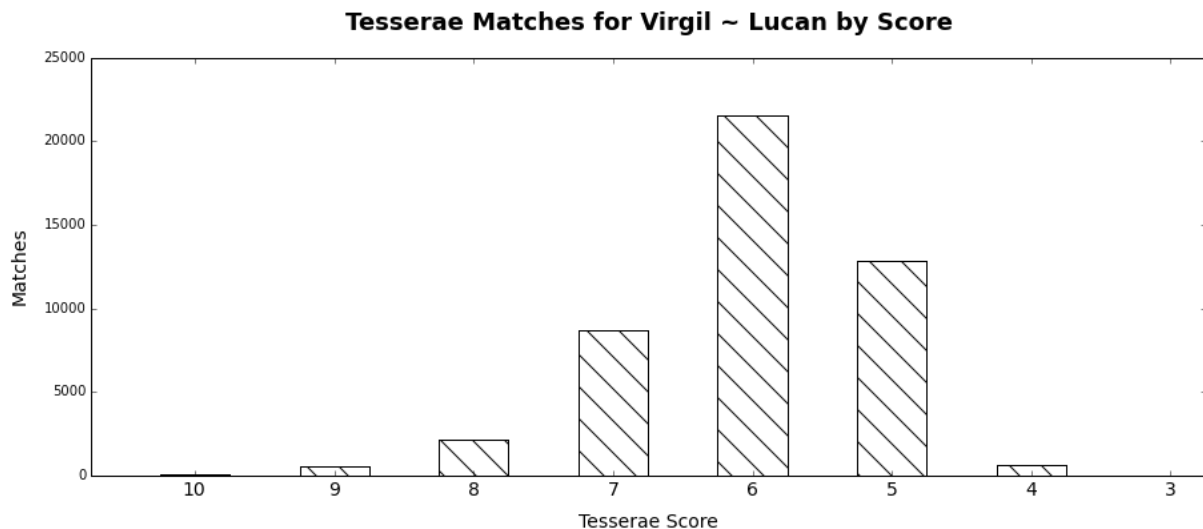


Figure 1. Number of Tesseræ matches by score for Virgil’s *Aeneid*.

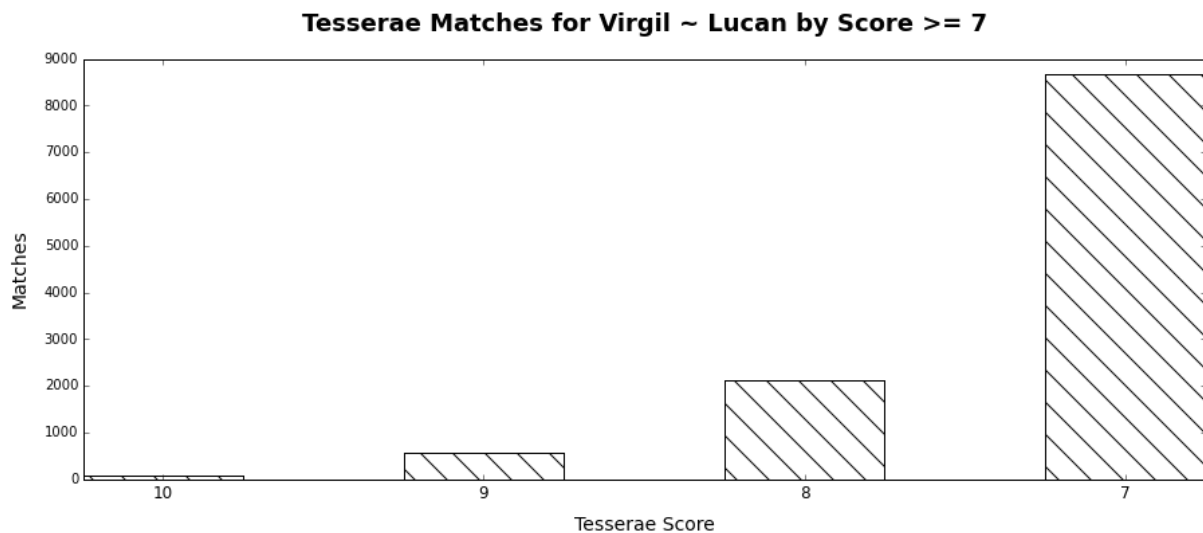


Figure 2. Number of Tesseræ matches by score for Virgil’s *Aeneid* at or above a threshold of 7.

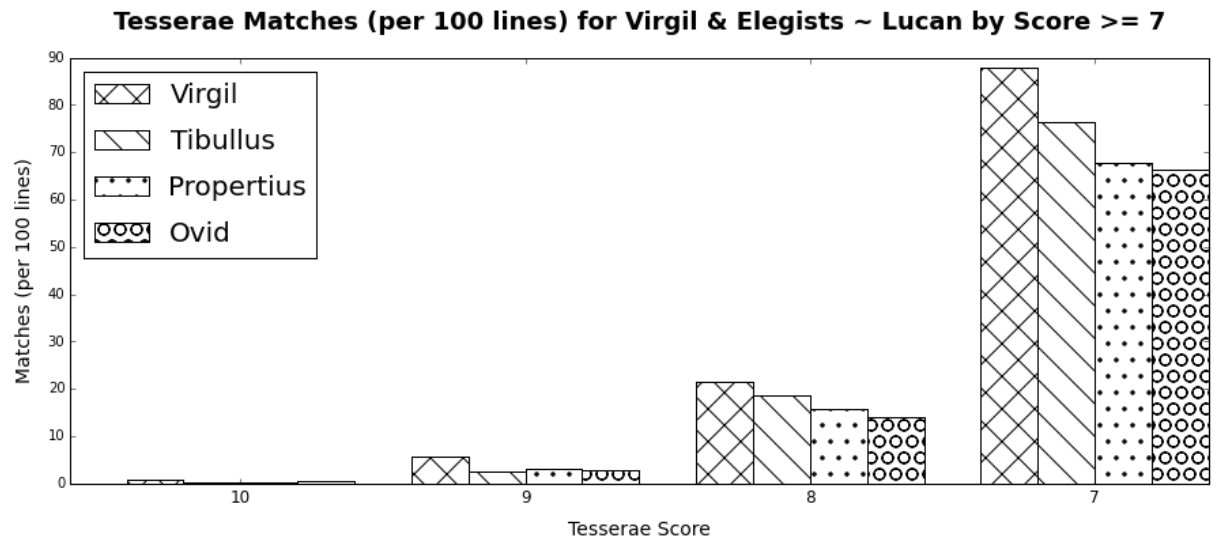


Figure 3. Chart showing the number of Tesseræ matches by score (at or above a threshold of 7) for the Latin love elegists and Virgil's *Aeneid* normalized to count per 100 lines as noted in Table 3.

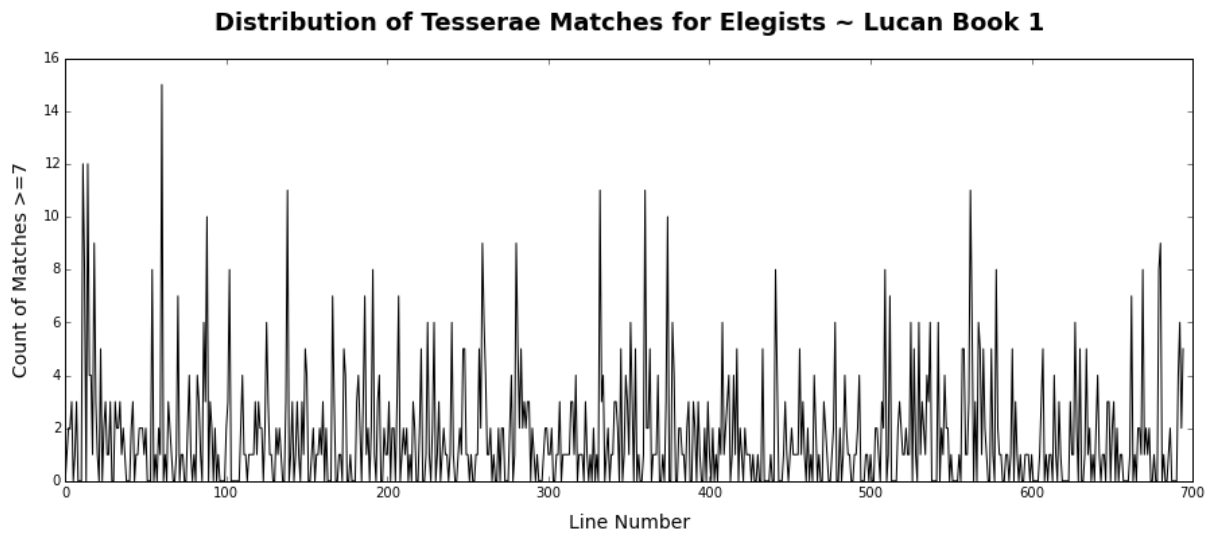
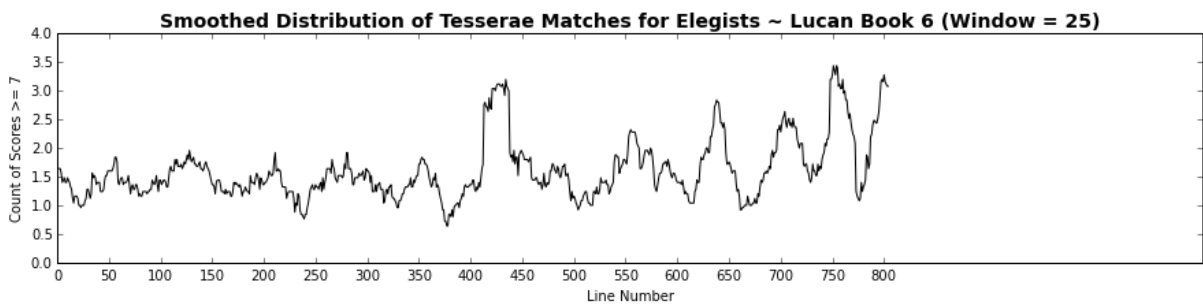
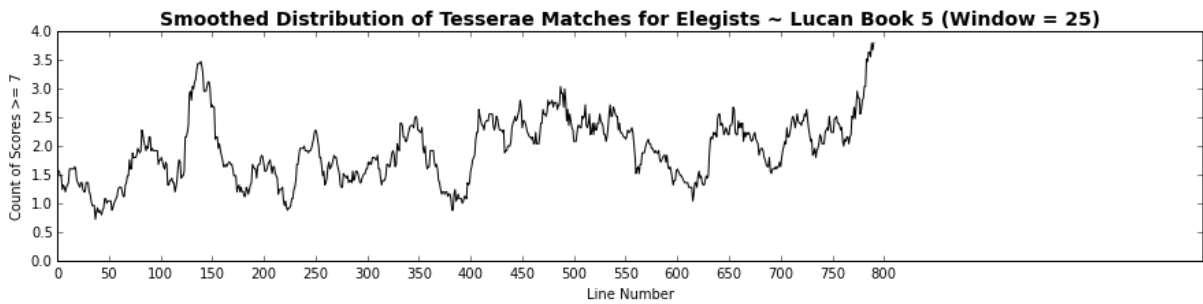
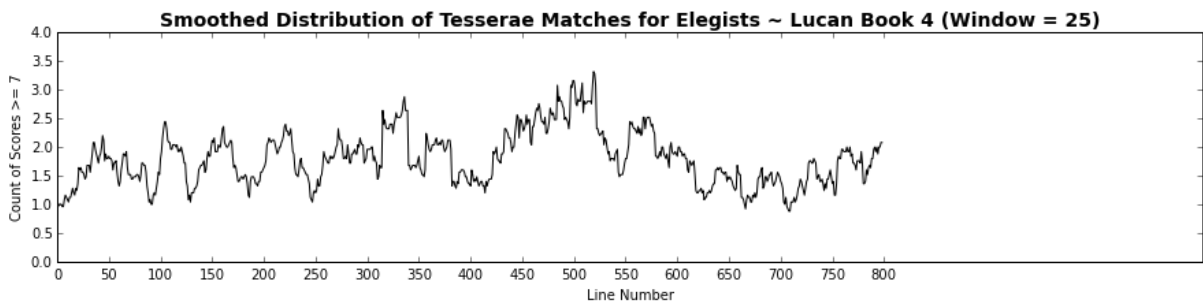
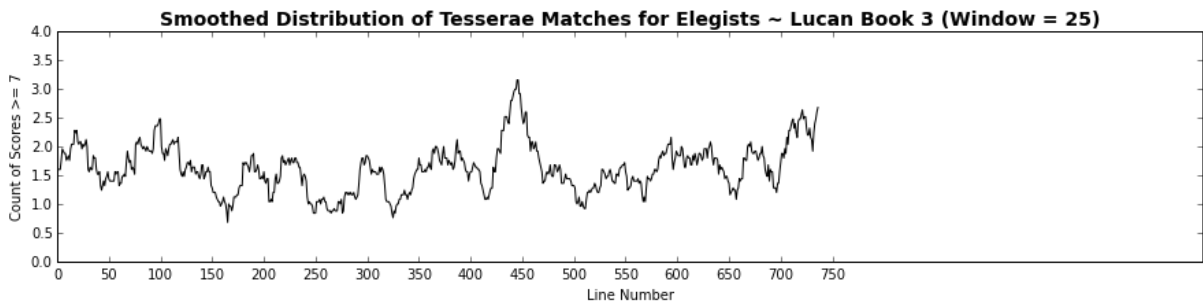
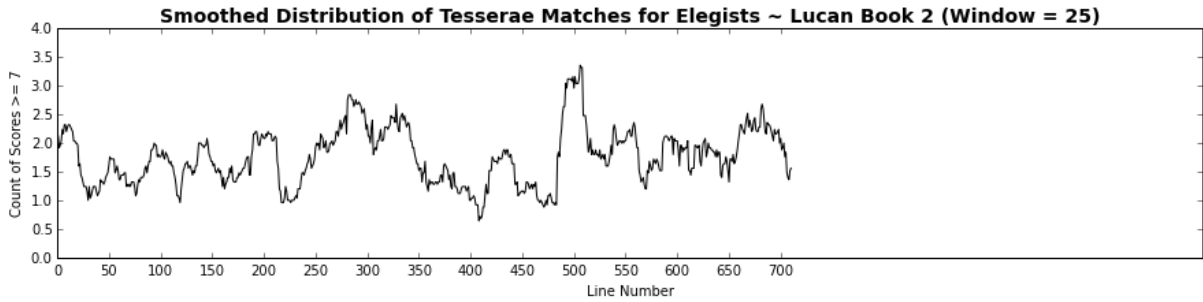
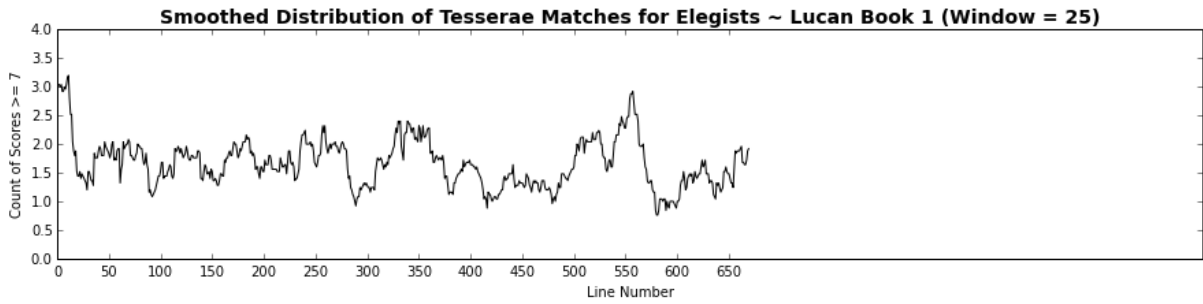


Figure 4. Total count of Tesseræ matches per line of book 1 of Lucan's *Bellum Civile* for the complete set of elegiac texts.



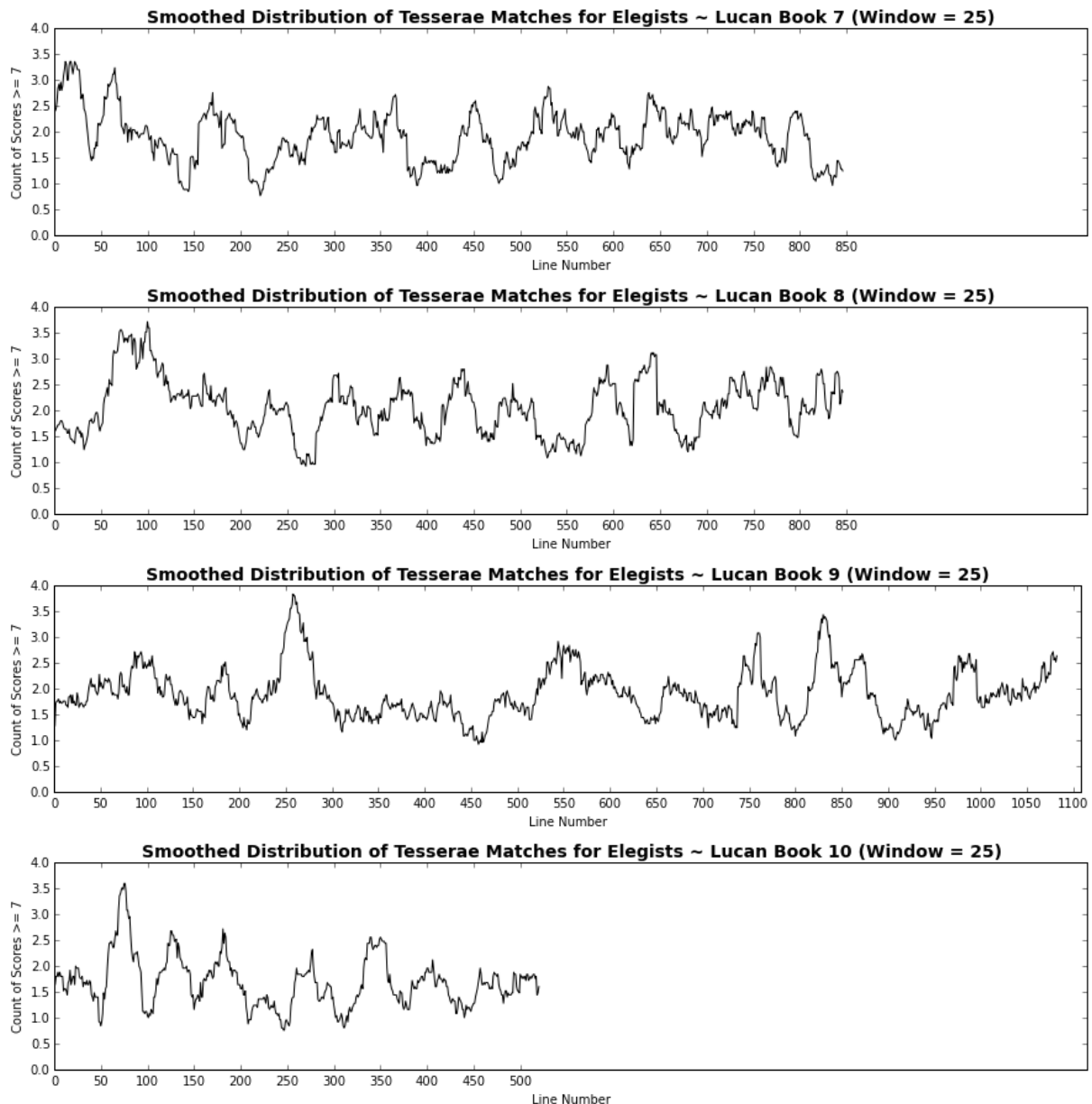


Figure 5. Average count of Tesseract matches for the elegists in the ten books of the *Bellum Civile*, smoothed using a window of 25 lines.

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