



Quantity or Stress? Sequential Analysis of Latin Prosody*

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ABSTRACT

Degrees of linear orderedness of Latin hexameter samples, coded respectively as sequences of long vs. short and stressed vs. unstressed syllables, were compared using the ARIMA method. Formal stochastic process models describing the sequential (metrical and/or rhythmical) text structure were identified and subsequently interpreted in linguistic terms. The average percentage of the original variance explained by the models turned out to be much higher for the stress series than for the series based on quantity coding. Contrary to received opinion, the underlying basis of Latin rhythm was not quantity but dynamic stress placed on the strong parts of consecutive metrical feet.

INTRODUCTION

An overview of a number of encyclopaedias and academic textbooks in linguistics indicates that the most frequently quoted example of a language with quantity-based prosody is Latin.¹ This kind of prosody is commonly contrasted with the tonality systems found in Asian languages as well as with dynamic expiratory stress, typical of most of the Indo-European languages. The function of quantity in Latin becomes prominent especially in those texts that comply with specifically Latin norms of versification. Acquaintance with Latin versification allows us to represent a text (now inescapably written) as a sequence of metrical feet composed of long and short syllables.

Clear as the picture seems to be, it remains slightly ambiguous. First, one cannot help asking why the vast majority of the Indo-European languages (especially the Romance languages derived from Latin) exhibit stress in the

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¹Cf. D. Crystal's remark in *The Cambridge Encyclopedia of Language*: 'By contrast, the length of a syllable (whether long or short) was a crucial feature of rhythm in Latin' (Crystal, 1997, p. 171).

advance is that both sequential orders, founded on syllable length and stress, respectively, play a part in determining text rhythm, the role of quantity being more prominent.

CORPUS AND QUANTIFICATION PROCEDURES

To verify the advanced hypothesis, we coded several samples of Latin hexameter by different authors (Horace—20, Ovid—10, Virgil—20).⁸ The average sample length was 150 syllables. Our choice of textual material is justified by the contents of the hypothesis to be tested. In spite of its limitations, hexameter leaves the writer some space for variation in shaping the rhythmical structure of verse. Consequently, the rhythmical structure of hexameter is neither fully predictable (as opposed to syllabic verse with a fixed distribution of stresses), nor completely devoid of any formal determinants of rhythm (as opposed to ordinary prose).

The purpose of quantifying the samples was to produce two corresponding sequences for each: a quantity-based sequence and a stress-based sequence. Using a numeral scale of the ordinal kind, we assigned value 1 to marked syllables (i.e., long or stressed, respectively) and value 0 to unmarked syllables (i.e., short or unstressed, respectively). Previous research has shown binary coding of this sort to be very effective in the sequential analysis of prosodic and metrical phenomena in Hebrew (Azar & Kedem, 1979) and English (Bratley & Ross, 1981). Sequential analyses of Polish (Pawłowski 1997, 1999) and Russian samples (Pawłowski 2000), based in like manner on the binary coding of stress, have proved very felicitous as well.

The method of quantification is exemplified by the treatment of the following distich (Verg. *Aen.* III 236–237):

*haud secus ac iussi faciunt tectosque per herbam
disponunt ensis et scuta latentia condunt.*

which can be represented as a sequence of metrical feet:⁹

d s d s d c
s s s d d s

⁸A detailed list of the texts under analysis is included in the appendix. It was not within the scope of our study to examine the samples for possible differences depending on authorship or interconnections between particular texts.

⁹Notations: d – dactyl, s – spondee, c – catalectic dactyl.

or long and short syllables:

_UU__UU__UU_U
____UU_UU__

or else stressed and unstressed syllables:

/UU/_/UU/_/UU/U
/_/_/_/UU/UU/_

RESEARCH METHOD

Given two samples {ABABABABAB...} and {BAAABAABBB...}, where A and B represent any relevant linguistic units, it would seem that in both samples elements A and B have the same positional parameters (e.g., frequency) and statistical distribution. From the linguistic point of view, however, both 'messages' are unquestionably different. A reliable description which takes into account, as a relevant feature, the order of units is feasible only with the help of sequential characteristics, such as the autocorrelation function (ACF), frequency spectrum, model of the underlying stochastic process, matrix of transition or conditional entropy. It should be stressed here that quantitative methods of sequential data analysis, both numerical (ARIMA) and probabilistic (Markovian and information theory models) allow us to describe any series of units, irrespective of their linguistic nature and/or method of coding (numeric or symbolic).

In the present study, discrete time-series, generated from Latin texts and representing sequences of long vs. short and accentuated vs. non-accentuated syllables, are treated by means of the ARIMA method in the time domain.¹⁰ The choice of the ARIMA method was determined by its efficiency in the analysis of discrete linguistic longitudinal data, as well as its efficiency in other branches of the human sciences.¹¹ It produced very good results in the

¹⁰The ARIMA method was described in detail in our earlier studies (Pawłowski, 1998) and will not be the subject of exposition here. The original source on the ARIMA method is Box and Jenkins's work (1976). For linguistic applications handbooks are recommended (Glass et al., 1975; Montgomery & Johnson, 1976, 188–240; McCleary & Hay, 1980; Nurius, 1983; Whiteley, 1980; Courtrot & Droesbeke, 1984, p. 67–76; Gottman, 1981; Stier, 1989).

¹¹'ARIMA modelling is perhaps the most commonly encountered and widely used of several stochastic process models adapted for use with time-series data.' (Nurius, 1983, p. 222).

| Quantity | Metrical stress |
|------------------|------------------|
| 111111110011 | 1010101010010 |
| 10011111110011 | 10010101010010 |
| 100111110010010 | 100101010010010 |
| 1111111110011 | 1010101010010 |
| 100111001110011 | 100101001010010 |
| 1001001001110011 | 1001001001010010 |
| 1111111110010 | 1010101010010 |
| 100111001110010 | 100101001010010 |
| 11111110010011 | 10101010010010 |
| 100111001110010 | 100101001010010 |

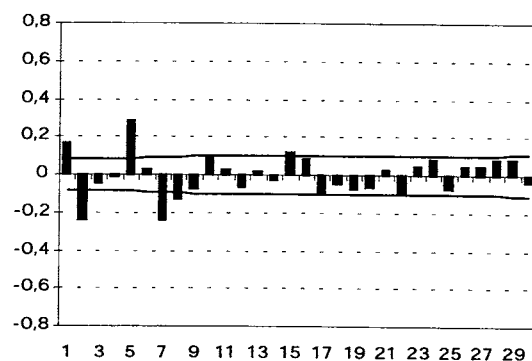


Fig. 1. ACF for the quantity series.

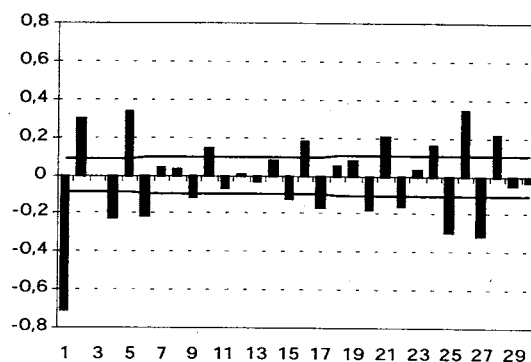


Fig. 2. ACF for the stress series.

The selection of an estimated model is influenced by the functions of ordinary partial autocorrelation. If we assume that the process is going to be of the simple type (AR or MA), not of the mixed kind (ARMA or SARMA), we can use the chart in Table 2 (Gottman 1984, p. 142).

It goes without saying that only seldom do real-life processes fully correspond to the idealized schema, the researcher often having to make intuitive decisions which are at least partly arbitrary. In this particular case, we decided to accept AR(5) or MA(2) as the most probable models for the

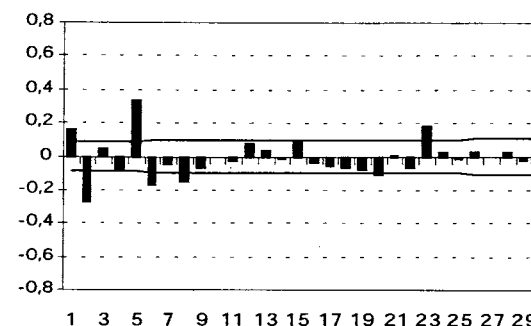


Fig. 3. PACF for the quantity series.

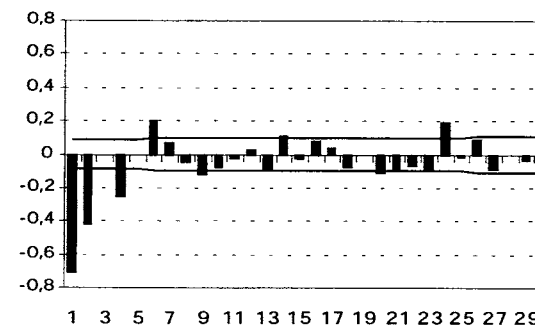


Fig. 4. PACF for the stress series.

Table 2.

| | ACF | PACF |
|-----------|------------------|------------------|
| AR(p) | Dies out | Stops at lag p |
| MA(q) | Stops at lag p | Dies out |

quantity sequence, and AR(2) or AR(4) for the stress sequence. As we have already claimed, the criterion of fit is the percentage of the original variance explained by the model (V_e). Table 3 contains percentage statistics on the extent to which the models adopted in our study match the data.

Table 3.

| Quantity | | Metrical stress | |
|--------------|--------------|-----------------|--------------|
| AR(5) 19% | MA(2) 15% | AR(2) 58% | AR(4) 61% |

The results for V_e corroborate our earlier conjectures. Although the quantity series does contain a deterministic component, it is only responsible for less than 20% of the total variance exhibited by the time series under analysis. Consequently, the quantity sequence is hardly predictable. A markedly different result is obtained when we analyse the stress sequence, whereby the adopted stochastic process models account for as much as 60% of series variance. It follows from this observation that, statistically considered, the stress sequence is more rhythmical and more predictable than the quantity sequence. Besides the percentage of the variance explained, model selection is also influenced by considerations of simplicity, or the number of parameters involved in the analysis. In general, a simpler model will usually be thought more parsimonious than a more complicated one and therefore superior to it. Guided by the principle of simplicity, we accepted as optimal the following models: MA(2) for the quantity sequence, and AR(2) for the stress sequence. The moving-average model for the quantity sequence could be defined as in formula (1):

$$x_t = e_t + 0,344e_{t-1} - 0,367e_{t-2} \quad (1)$$

where x_t —series value at moment or position t ; e_t —value of normally distributed noise $N(0,1)$ at moment or position t ; while the autoregression model for the stress sequence could be represented as in formula (2):

$$x_t = -x_{t-1} + 0,416x_{t-2} + e_t \quad (2)$$

where x_t —series value at moment or position t ; e_t —value of normally distributed noise $N(0,1)$ at moment or position t .

It is noteworthy how profoundly different the two stochastic models are from each other despite their being estimated for essentially the same stretch of text. This remarkable difference provides us with further evidence for the claim that quantity- and stress-based rhythmical patterns are in fact unrelated.

The crucial, and arguably most difficult, task in the quantitative analysis of textual data is to convincingly interpret mathematical models in linguistic

terms. The linguistic counterpart of parameter x_{t-1} is of course the syllable feature (0—short or unstressed, 1—long or stressed). The coefficients in the two models (1 and 2) indicate the degree of correlation between the feature of a given element and the features of the preceding elements in the series. Autocorrelation values can be interpreted in like manner. If the coefficient of x_{t-1} approaches -1 , for example, the correlation is saliently negative, suggesting that in most cases a stressed syllable will tend to 'force' an unstressed one as the immediately succeeding element in the series, and vice versa. This can be seen in the stress-based model (see equation 2). The seemingly strange ACF for the quantity series (Figure 1), and the salient bar for lag 5 ($r_5 = 0.288$) in particular, indicate a noticeable positive correlation between syllable features at t and $t-5$ in the linear order of text. Taking a closer look at the quantity series will enable us to see multiple-syllable sequences of long syllables divided by two- or one-syllable sequences of short syllables.

If we restrict our analysis to test the hypothesis as outlined above, however, an equally meticulous analysis of the particular samples will turn out to be unnecessary as a means of arriving at generalizations. In that case, we will need a more comprehensive parameter to capture the phenomenon of text rhythm. As stated above, the percentage of the original variance explained by the model (V_e) meets this condition for possible criteria. The parameter in question shows the degree of sequential orderedness characteristic of a given time-series under analysis, and in linguistic terms it constitutes a comprehensive measure of text rhythm.¹⁴ The value of V_e can be computed for any series, which clearly means that it is not dependent on the kind of textual input fed into the algorithm. A practical gain from using this method is that it enables the researcher to compare literally any two samples, irrespective of their language, style (verse or prose) or length. It was precisely the use of V_e that made it possible for us to falsify our initial hypothesis.

SUMMARY RESULTS

In analysing the samples, we found it rather difficult to identify a single distinctively recurring pattern of autocorrelation. For this reason, it seemed rational to us to compute the average ACF values for lags 1–5 for all the samples under analysis. Although no exactly determined confidence intervals

¹⁴On no account should 'the degree of orderedness' be automatically interpreted in aesthetic or axiological terms. In the present context 'orderedness' is merely a typological category.

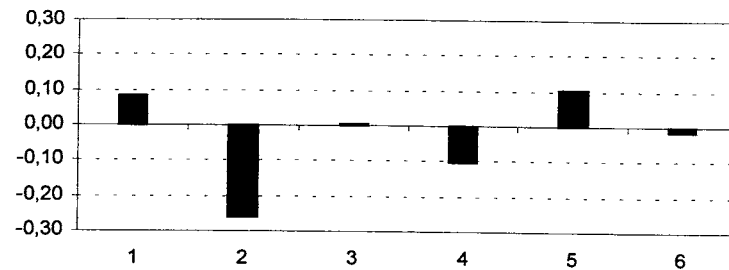


Fig. 5. Average ACF for the quantity sequence.

emerge from the computation, one can note that only r_2 is statistically significant, while the other values either verge on non-significance (lags 1, 4, 5) or simply equals zero (lags 3 and 6). This implies a relatively minor degree of linear arrangement of long and short syllables in classical hexameter. An analogous autocorrelation graph for the stress series proved unnecessary. The model presented in the previous chapter, indicating a strong negative correlation for consecutive syllables, recurred with remarkable regularity in all the samples.

A more accurate verification of the hypothesis that hexametric rhythm is generated by both quantity and stress, however, was made possible by comparing the average values of the comprehensive parameter V_e for each sequence (as stated above, V_e constituted a numerical measure of a text's sequential orderedness). The results we obtained have failed to corroborate the initial hypothesis, which predicted that the respective degrees of rhythmical order generated by the quantity and stress sequences should be rather high and mutually comparable. The value of V_e averaged 15% for quantity sequences, and 61% for sequences based on dynamic stress. Contrary to widespread opinion, it is not the length of syllables but the regularity of stress in consecutive metrical feet that produces the rhythmical fabric of hexameter. In more general terms, it may be claimed that, recited with metrical stress, Latin texts will tend to sound more rhythmical than is the case when only quantity or both quantity and metrical stress play a role in oral performance.

DISCUSSION

It has been shown that, in the case of Latin hexameter, quantity didn't directly underlie text rhythm because, as for the linear arrangement of text, long and

short syllables correlated too poorly within the sequence. If we still wanted to defend quantity's decisive contribution to text rhythm, it could only be by dispensing with our original definition of rhythm altogether. The fact that our finding stands in certain contradiction to a time-honoured theory calls for serious reflection and careful interpretation. The phrase 'certain contradiction' is not a mere euphemism here. It is surely undeniable that classical Latin versification was based on the principle of quantity, the phenomenon of metrical stress being recognised as a factor in its own right. But on the other hand, the glaring disproportion between the V_e values for quantity and stress is a hard, empirical fact and, as such, cannot be simply ignored.

Our conjecture is that, in the case of classical Latin, quantity was a superstructural form, a cultural addition, as it were, that no doubt organised Latin versification and was related to the musical context of performance, peculiar to certain texts.¹⁵ However, the sequential structure based on quantity apparently failed to overlap with the natural rhythmical patterns that facilitate the generation, understanding, and memorisation of texts under any communicational circumstances.¹⁶ Such rhythmical patterns, described by means of formal models, have been shown to function effectively in the prosody of modern languages (cf. Dreher et al., 1969; Pawłowski, 1997, 1999, 2000), and much evidence supports the claim that they may well be a universal property of natural language. Disruptive as our conclusion seems to previous accounts of the subject, it presents Latin intractably as no exception to this general rule.

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¹⁵In the course of reciting a text, the performer would signal the strong part of a metrical foot (*thesis* or *arsis*) by pounding his foot against the ground or producing extra sounds with a musical instrument.

¹⁶Rhythm is one of the mechanisms often used in mnemonic techniques.

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APPENDIX

Hor. *Serm.* I 3, 66–75; I 4, 115–124; I 9, 63–72; I 10, 41–50; II 3, 176–185; II 4, 4–13; II 5, 28–37; II 7, 108–117; II 8, 71–80; Hor. *Epist.* I 1, 53–62; I 7, 1–10; I 11, 19–28; I 13, 10–19; I 16, 24–33; I 20, 4–13; II 1, 199–208; Hor. *Ars* 60–69; 147–156; 220–229; 351–360; Verg. *Aen.* I 419–428; III 229–238; IV 634–643; V 72–81; VI 585–594; VII 770–779; IX 88–97; X 668–677; XI 764–773; XII 233–242; Verg. *Georg.* I 84–93; I 328–337; II 184–193; III 16–25; III 414–423; IV 206–215; Verg. *Ecl.* I 59–68; IV 11–20; VII 57–66; IX 32–41; Ov. *Met.* I 454–463; II 52–61; III 583–592; IV 742–751; V 228–237; VI 503–512; VII 147–156; VIII 713–722; IX 370–379; X 702–711.