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REVISED DOCUMENT ON<br>ANALYZING VISUALLY OBSERVED DATA IN TWO GRASS SPECIES

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## Analysing visually observed data in Dactylus and Festuca

## 1. Introduction

At the XIV-th meeting of the TWC in Hannover threshold models were introduced as a means to analyse visually observed data. In TWC/14/12 a brief exposition of theory was given together with two examples. Those examples did not cover routine application of threshold models to assess distinctness and uniformity. In this paper experiences with threshold models are reported for the analysis of the character which is called alternativité in French for the grasses cocksfoot (Dactylus) and tall fescue (Festuca élevé). Special attention is given to checking the assumption of uni-modality for the threshold model. To comply with this assumption observed categories may have to be combined. An alternative to the threshold model that is close to a standard COY-D is presented, and was found to be useful as a conservative, i.e., less powerful, approximation.

## 2. Summary of theory and assumptions

As explained in TWC/14/12 the key idea of a threshold model is that the observed category scores, $\mathrm{y}_{\mathrm{i}}$, for a treatment (variety) i are the expression of a continuous underlying variable, $\mathrm{U}_{\mathrm{i}}$, that cannot be observed and that follows a certain distribution like the normal or the logistic. The underlying variable $\mathrm{U}_{\mathrm{i}}$ determines the category in which observations on the treatment can lie. The categories 1 to $C$ are separated by so-called cut points or thresholds: $\square$ to $\square_{-1}$. If the underlying variable lies in the interval before the first threshold, $\square$, the observation belongs to the lowest category 1, if the underlying variable lies in the interval $\square$-1 till $\square$, the observation belongs to category c ( $c=2 \ldots \mathrm{C}-1$ ), and finally above the last threshold, $\square_{-1}$, the observation belongs to the highest category $C$. The underlying variable $U_{i}$ is a theoretical construct that allows us to model the observed category scores. A popular type of threshold model is the so-called proportional odds model. This model has the form $\log \left(\gamma_{i c} /\left(1-\gamma_{i c}\right)\right)=\theta_{c}-\sum \beta_{j} x_{i j}$, where $\square_{i c}$ is the probability that the observation lies in one of the categories 1 to c. The distribution of $\square_{c}$ can be obtained from the multinomial. The parameter $\square_{j}$ stands for the regression coefficient to the explanatory variable $x_{j}$ (varieties in our case). An important assumption to be fullfilled for threshold models is that the underlying variable is uni-modal. Roughly said this means that a histogram of the numbers of observations in the different categories (for a particular variety) should have more observations in the middle categories than in the extreme categories. When this is not the case categories may be combined to see whether this solves the problem of multi-modality. For the application of the proportional odds form of the threshold model it is also necessary that the difference between treatments (varieties) on the log-odds scale, $\log \left(\square_{i} /\left(1-\square_{i}\right)\right.$, is independent of the category c .

## 3. Dactylus and Festuca data description and rearrangement

Analysed were data from Dactylus and Festuca, covering the years 1995 to 1997. Per year a trial in three replicates was done, where each replicate contained 20 individual plants. Category scores were given to individual plants. The totals of observed category scores over years and replicates per variety are given for Dactylis in Table 1 and for Festuca in Table 7. Looking at these tables clearly the assumption of uni-modality is not fulfilled for neither of both species. For Dactylis most observations occurred in the extreme categories of 2 and 8, with some observations in between. For Festuca most observations occurred in the categories 2 and 6 with a dip in the categories 3 and 4 . For both sets of data rearrangement of categories was a prerequisite for application of a threshold model.

For Dactylis category 2 became one category and all other categories were combined to form a second category (Table 2). Effectively, the character alternativité becomes a binary trait, either category 2 is observed or a higher category. After this rearrangement a threshold model would again be feasible. The threshold model gets a very simple structure with only two categories. The unique threshold $\square$ is automatically fixed at 0 and does not need to be estimated. The proportional odds model becomes a standard general linear model with logit link and binomial distribution, a form known under the name of logistic regression: $\log \left(\gamma_{i c} /\left(1-\gamma_{i c}\right)\right)=\sum \beta_{j} x_{i j}$ (McCullagh and Nelder, 1989).

For Festuca three categories could be formed by combining categories 1 and 2 to a new category 1 , fusing 3,4 and 6 to a new category 2 , while category 8 became the new category 3 (Table 8). Because after rearrangement so few categories remained, the possibility for testing simultaneously distinctness and uniformity was discarded. Testing of differences in uniformity, differences in width of the underlying dsitributions, makes only sense when more than 3 categories are retained.

## 4. Analysis of variance and threshold model

As discussed in TWC/14/12, for the analysis of visually observed data threshold models would be the most appropriate models. Thus, the Dactylis and Festuca data were analysed with threshold models. After presentation of TWC/14/12 questions arose to simpler alternatives to the threshold model. One alternative to the application of threshold models was investigated. First acknowledge that the proportional odds model can be interpreted as a series of logistic regressions for the cumulative category probabilities, $\square_{i}$, for $\mathrm{c}=1 \ldots \mathrm{C}-1$. As there were only two categories for Dactylis, the threshold model was equivalent to a logistic regression. For Festuca the proportional odds model could be approximated by logistic regressions for $\Pi_{1}$ and $Z_{i}$. Next, an approximation to a logistic regression is ordinary regression on the logit transformed cumulative category counts, $\log \left(y_{i c} /\left(y_{i c}-y_{i c}\right)\right.$, which is equivalent to an analysis of variance for our type of data. The residuals of the logistic regressions for $\square_{1}$ and $\square_{2}$, and to a lesser degree the residuals of the corresponding regressions on the logit transforms, can be used as checks for the threshold model.

Thus results obtained by applying threshold models may be compared to those of analysis of variance on logit transformed cumulative category counts. For calculating pairwise differences following standard COY-D practices, the Variety by Year interaction mean square
serves as the basis for the calculation of a standard error of difference between varieties. For generalized linear models, including threshold models, the equivalents of mean squares and sums of squares are so-called mean deviances and deviances. The latter may be treated as if they were mean squares and sums of squares. Therefore, for assessing distinctness in threshold models, the Variety by Year mean deviance was used to calculate the standard errors of differences. In contrast to the standard COY-D, in threshold models every comparison has its own standard error.

For Dactylis distinctness was assessed by a logistic regression model, after which all pairwise differences were calculated. For comparison a classical COY-D procedure was followed for transformed category counts. Both types of comparison were done at $\mathrm{p}<0.01$.

For Festuca various threshold models were fitted and pairwise comparisons were made. For comparison COY-D was applied to the logits of the cumulative category counts for the classes 1 and 2.

## 5. Results Dactylis

In Table 3 it is shown how many times distinctness was assessed for each variety using the different methods. As can be seen the results of COY-D (AOV) and the threshold model were in good agreement. Every variety could be identified as being distinct from at least 6 others by COY-D. For the threshold model this figure was 8 . The analyses of variance and deviance are given in Tables 4 and 5 . As can be seen both types of analysis indicate the presence of Variety by Year interaction. This interaction may be tried to be reduced by Modified Joint Regression to achieve a more powerful test on distinctness. The analysis of deviance table for the threshold model allows the same kind of inference as the analysis of variance table in the application of COY-D. Table 6 shows in detail all incidences of distinctness as assessed by the threshold model.

## 6. Results Festuca

A first threshold model was fitted to the Variety by Year table of category counts (Table 9). The Variety by Year interaction mean deviance then was used to calculate standard errors of differences at $\mathrm{p}<0.01$. Results per variety for the threshold model can be found in Table 10 (TM), at least 7 varieties were found to be distinct from each individual variety. The results of the COY-D procedures on the cumulative logit transforms can be seen in Table 10 in the columns AOV1, AOV2 and AOV1-2. The last column is of special interest as its shows the consistency of the two COY-D's, only distinctnesses which were distinct for both cumulative logits at $\mathrm{p}<0.001$ are present in the counts of the column AOV1-2. The disctinctnesses of the simultaneous COY-D at $\mathrm{p}<0.001$ were almost completely contained in those of the threshold model at $\mathrm{p}<0.01$ (AOV-TM, Table 10). The simultaneous COY-D behaved like a conservative approximation to the threshold model, because the threshold model allowed 802 pairwise comparisons to be distinct against 476 for the simultaneous COY-D. Nevertheless, the behaviour of the simultaneous COY-D shows that it might be an alternative in cases where software for threshold models is less accessible. Increasing the COY-D p-value led to both more distinctnesses which were shared with the threshold model as to distinctnesses that were
only found distinct by the COY-D procedure and not by the threshold model. These latter distinctnesses do not seem to be very trustworthy.

In addition to the use of the logit transforms for the COY-D procedure, these transforms can also be used to check the assumption of constant differences between varieties, i.e., the difference should be independent of the category. To check the assumption one can regress one logit on the other. If the slope does not deviate from 1, the assumption is fulfilled. For the Festuca data the slope measured 1.16 with an $95 \%$ confidence interval of $+/-0.19$, while the fit was good, $r^{2}=0.64$. Figure 1 shows a graph of the cumulative logit for category 2 against that for 1 .

The residuals from the analyses of variance on the cumulative logits are useful to check for anomalies in the data. Suspect observations are likely to cause problems also in the threshold model. Figures 2 a and 2 b contain half-normal plots of the residuals for the cumulative logits of category 1 and 2. There is one clearly deviating observation in Figure 2a. It was identified as being due to the remarkable behaviour of variety 416530 in 1996 (Table 9). For the remainder, the half-normal plots gave no reason for concern as the residuals behaved as they should for normally distributed observations, namely to lie on a straight line through the origin.

For determining whether Variety by Year interaction was present a threshold model was fitted containing a so-called mixed model for the underlying variable $\mathrm{U}_{\mathrm{i}}$, with as fixed terms Variety, Year and Variety by Year interaction, and as random terms Replicates within Years and Plants within Replicates within Years. The most important result from this exercise was that there appeared to be substantial interaction. To investigate whether the Variety by Year mean deviance might be reduced by modified joint regression the estimated Variety by Year means for the underlying variable $U$ were regressed by ordinary regression on the estimated Variety main effects for each of the years 1995, 1996 and 1997. Results are shown in Figures 3a, b and c and Table 11. There were clear differences between the years, with 1995 reducing differences (slope 0.73), 1996 increasing differences (slope 1.34), and 1997 being more or less average (slope 0.93). Extending threshold models by a formulation for modified joint regression thus seems worthwhile.

## 7. Conclusions

The application of threshold models to assess distinctness in grasses led to encouraging results. Using a threshold model all varieties of Dactylis and Festuca could be distinguished on the basis of the visually assessed character alternativité. Before application of threshold models the assumption of uni-modality of the underlying response should be checked. It wil often be necessary to combine observation categories to comply with this requirement. Another assumption to be checked for the proportional odds model is whether the difference between varieties is constant over the categories. As an alternative to the threshold model COY-D like procedures can be used on the logits of the cumulative category counts. For this alternative to work, p -values should be smaller than the standard UPOV values and only consistent distinctnesses over the various categories should be taken into account.

## Acknowledgement

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## Reference

McCullagh, P. \& J.A. Nelder (1989) Generalized linear models, 2nd. edn., Chapman and Hall.
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Table 1. Observed category scores for alternativit in Dactylis (Cocksfoot).

| Number | 2 | 4 | 6 | 7 | 8 Total |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |
| 423320 | 157 | 1 | 2 | 0 | 20 | 180 |
| 426850 | 147 | 2 | 8 | 0 | 23 | 180 |
| 508850 | 173 | 1 | 1 | 0 | 5 | 180 |
| 516690 | 170 | 1 | 5 | 0 | 4 | 180 |
| 517770 | 123 | 15 | 15 | 0 | 27 | 180 |
| 520070 | 135 | 4 | 7 | 0 | 34 | 180 |
| 520080 | 159 | 2 | 11 | 0 | 8 | 180 |
| 521780 | 165 | 6 | 7 | 0 | 2 | 180 |
| 530200 | 125 | 7 | 15 | 0 | 33 | 180 |
| 559770 | 153 | 2 | 7 | 0 | 18 | 180 |
| 566170 | 132 | 4 | 12 | 0 | 32 | 180 |
| 581380 | 161 | 1 | 7 | 0 | 11 | 180 |
| 599260 | 132 | 2 | 13 | 0 | 33 | 180 |
| 609050 | 148 | 4 | 7 | 0 | 21 | 180 |
| 620150 | 170 | 1 | 2 | 0 | 7 | 180 |
| 620730 | 171 | 0 | 3 | 0 | 6 | 180 |
| 655360 | 1 | 2 | 35 | 0 | 142 | 180 |
| 655650 | 162 | 5 | 3 | 0 | 10 | 180 |
| 655870 | 168 | 2 | 3 | 0 | 7 | 180 |
| 655880 | 151 | 5 | 4 | 0 | 20 | 180 |
| 667410 | 174 | 3 | 0 | 0 | 3 | 180 |
| 678290 | 167 | 2 | 4 | 0 | 7 | 180 |
| 678430 | 117 | 4 | 11 | 1 | 47 | 180 |
| 678580 | 150 | $\bigcirc$ | 7 | 0 | 23 | 180 |
| 678920 | 117 | 24 | 27 | 0 | 12 | 180 |
| 903120 | 140 | 3 | 13 | 0 | 24 | 180 |
| 903150 | 143 | 10 | 8 | 0 | 19 | 180 |
| 1127680 | 118 | 5 | 18 | 0 | 39 | 180 |
| 1348090 | 73 | 14 | 22 | 0 | 71 | 180 |
| 1371510 | 148 | 6 | 9 | 0 | 17 | 180 |
| Total | 4250 | 138 | 286 | 1 | 725 | 5400 |


| Table 2. Rearranged category scores for Dactylis. |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| 1 | 423320 | DP 6502 | 157 | 23 |
| 2 | 426850 | AND 687 | 147 | 33 |
| 3 | 508850 | LUCYLE | 173 | 7 |
| 4 | 516690 | LULLY | 170 | 10 |
| 5 | 517770 | DORISE | 123 | 57 |
| 6 | 520070 | LUDE | 135 | 45 |
| 7 | 520080 | LUTETIA | 159 | 21 |
| 8 | 521780 | MODAC | 165 | 15 |
| 9 | 530200 | CAMBRIA | 125 | 55 |
| 10 | 559770 | ARLY | 153 | 27 |
| 11 | 566170 | AMPLY | 132 | 48 |
| 12 | 581380 | FURLY | 161 | 19 |
| 13 | 599260 | ATHOS | 132 | 48 |
| 14 | 609050 | PORTHOS | 148 | 32 |
| 15 | 620150 | KID | 170 | 10 |
| 16 | 620730 | LUPRE | 171 | 9 |
| 17 | 655360 | MATOP | 1 | 179 |
| 18 | 655650 | SABORTO | 162 | 18 |
| 19 | 655870 | STARLY | 168 | 12 |
| 20 | 655880 | ACCORD | 151 | 29 |
| 21 | 667410 | BAR DGL 38 | 174 | 6 |
| 22 | 678290 | MOM DAC 17 | 167 | 13 |
| 23 | 678430 | 87-2 | 117 | 63 |
| 24 | 678580 | L-DGL 258 | 150 | 30 |
| 25 | 678920 | K 2 M | 117 | 63 |
| 26 | 903120 | FLOREAL | 140 | 40 |
| 27 | 903150 | PRAIRIAL | 143 | 37 |
| 28 | 1127680 | BARAULA | 118 | 62 |
| 29 | 1348090 | MOBITE | 73 | 107 |
| 30 | 1371510 | AS 26 | 148 | 32 |

Table 3.Number of Dactylis varieties that were found distinct at $p<0.01$ from a specific variety by an analysis of variance on logits (AOV) and by a generalized linear model with logit link and binomial distribution (GLM), plus the number of coincidences on distinctness. (* Total is sum column divided by 2).

| number | AOV | GLM | Coinc. |
| :---: | :---: | :---: | :---: |
| 423320 | 10 | 9 | 9 |
| 426850 | 6 | 10 | 6 |
| 508850 | 17 | 18 | 17 |
| 516690 | 14 | 15 | 14 |
| 517770 | 14 | 17 | 14 |
| 520070 | 13 | 12 | 11 |
| 520080 | 15 | 9 | 9 |
| 521780 | 10 | 11 | 10 |
| 530200 | 14 | 15 | 14 |
| 559770 | 7 | 9 | 7 |
| 566170 | 13 | 14 | 13 |
| 581380 | 12 | 10 | 10 |
| 599260 | 14 | 14 | 14 |
| 609050 | 8 | 10 | 7 |
| 620150 | 15 | 15 | 14 |
| 620730 | 14 | 16 | 14 |
| 655360 | 29 | 29 | 29 |
| 655650 | 8 | 10 | 8 |
| 655870 | 12 | 12 | 12 |
| 655880 | 6 | 8 | 4 |
| 667410 | 17 | 18 | 17 |
| 678290 | 11 | 12 | 11 |
| 678430 | 15 | 20 | 15 |
| 678580 | 6 | 9 | 6 |
| 678920 | 16 | 20 | 16 |
| 903120 | 11 | 10 | 9 |
| 903150 | 10 | 9 | 8 |
| 1127680 | 16 | 20 | 16 |
| 1348090 | 25 | 29 | 25 |
| 1371510 | 8 | 10 | 7 |
| Total* | 193 | 210 | 183 |

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Table 4. Analysis of variance for logit transformed Dactlyis data.

Variate: Alternativit

Source of variation d.f

Year.Rep stratum

Year
Residual
year.rep.*Units* stratum
Cult
29
Cult.Year
58
Residual
174
Total
269
466.0963
45.5479
68.1044

| 16.0723 | 41.06 |
| ---: | ---: |
| 0.7853 | 2.01 |
| 0.3914 |  |

0.7853
. 01
$46.0916 \quad 36.00$
1.2804
3.27
92.1832
7.6822
0.3914
m.s. v.r.
679.6141

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Table 5. Analysis of deviance for generalized linear model with logit link and binomial distribution for Dactlyis data.

```
*** Accumulated analysis of deviance ***
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|  |  |  | mean <br> deviance <br> ratio |  |
| :--- | ---: | ---: | ---: | ---: |
| Year | d.f. | deviance | deviance |  |
| Year.Rep | 2 | 195.4658 | 97.7329 | 65.65 |
| Cult | 6 | 22.0972 | 3.6829 | 2.48 |
| Cult.Year | 29 | 1154.2227 | 39.8008 | 26.74 |
| Cult.Year.Rep | 58 | 113.1846 | 1.9515 | 1.95 |
| Residual | 174 | 191.7480 | 1.1020 | 0.75 |
| Total | 5130 | 3916.1265 | 0.7634 |  |
|  |  |  |  |  |
|  | 5399 | 5592.8447 | 1.0359 |  |

Table 6. Distinctness at $p<0.01$ for Dactylis. $1=$ row number had significantly higher alternativit $\square$ than column number, $-1=$ vice versa, $0=$ no significant difference found.


| Table 7. Observed category scores for alternativit $\square$ in Festuca. Category |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Number |  |  |  |  |  |  |  |
|  | 407030 | 0 | 55 | 0 | 19 | 65 | 41 | 180 |
|  | 411190 | 0 | 52 | 0 | 16 | 71 | 41 | 180 |
|  | 416520 | 0 | 79 | 0 | 15 | 74 | 12 | 180 |
|  | 416530 | 0 | 61 | 1 | 9 | 83 | 26 | 180 |
|  | 416800 | 0 | 56 | 0 | 9 | 77 | 38 | 180 |
|  | 419770 | 0 | 77 | 0 | 15 | 76 | 12 | 180 |
|  | 423710 | 0 | 54 | 0 | 12 | 70 | 44 | 180 |
|  | 423810 | 0 | 86 | 0 | 16 | 69 | 9 | 180 |
|  | 426950 | 0 | 141 | $\bigcirc$ | 15 | 20 | 4 | 180 |
|  | 426960 | 0 | 113 | 0 | 9 | 47 | 11 | 180 |
|  | 512670 | 0 | 102 | 0 | 12 | 57 | 9 | 180 |
|  | 525950 | 0 | 88 | $\bigcirc$ | 14 | 72 | 6 | 180 |
|  | 526540 | 0 | 130 | $\bigcirc$ | 6 | 37 | 7 | 180 |
|  | 533760 | 0 | 113 | 0 | 16 | 47 | 4 | 180 |
|  | 533940 | 0 | 117 | $\bigcirc$ | 17 | 41 | 5 | 180 |
|  | 534010 | 0 | 47 | $\bigcirc$ | 19 | 84 | 30 | 180 |
|  | 538910 | 0 | 35 | $\bigcirc$ | 10 | 77 | 58 | 180 |
|  | 539130 | 0 | 101 | 0 | 14 | 55 | 10 | 180 |
|  | 539310 | 0 | 20 | 0 | 13 | 88 | 59 | 180 |
|  | 548500 | 0 | 83 | 0 | 17 | 74 | 6 | 180 |
|  | 552610 | 0 | 47 | 0 | 24 | 69 | 40 | 180 |
|  | 553110 | 0 | 73 | 0 | 10 | 92 | 5 | 180 |
|  | 559300 | 0 | 54 | 0 | 9 | 94 | 23 | 180 |
|  | 559340 | 0 | 36 | 0 | 11 | 102 | 31 | 180 |
|  | 559780 | 0 | 104 | $\bigcirc$ | 15 | 55 | 6 | 180 |
|  | 560280 | 0 | 33 | 0 | 14 | 83 | 50 | 180 |
|  | 566010 | 0 | 51 | 0 | 18 | 83 | 28 | 180 |
|  | 566270 | 0 | 46 | 0 | 17 | 100 | 17 | 180 |
|  | 566440 | 0 | 92 | $\bigcirc$ | 14 | 58 | 16 | 180 |
|  | 572470 | 0 | 43 | 0 | 13 | 75 | 49 | 180 |
|  | 572550 | 0 | 45 | 0 | 17 | 96 | 22 | 180 |
|  | 572590 | 0 | 48 | 0 | 9 | 85 | 38 | 180 |
|  | 580680 | 0 | 34 | $\bigcirc$ | 14 | 86 | 46 | 180 |
|  | 580770 | 0 | 85 | $\bigcirc$ | 15 | 63 | 17 | 180 |
|  | 581530 | 0 | 84 | 0 | 6 | 34 | 56 | 180 |
|  | 589890 | 0 | 33 | $\bigcirc$ | 14 | 92 | 41 | 180 |
|  | 589970 | 0 | 13 | 0 | $\bigcirc$ | 37 | 130 | 180 |
|  | 590060 | 0 | 54 | $\bigcirc$ | 14 | 82 | 30 | 180 |
|  | 598740 | 0 | 13 | 0 | 6 | 103 | 58 | 180 |
|  | 599220 | 0 | 51 | 0 | 23 | 81 | 25 | 180 |


| 599590 | 0 | 44 | 0 | 11 | 78 | 47 | 180 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 609060 | 0 | 49 | 0 | 11 | 83 | 37 | 180 |
| 609300 | 0 | 137 | 0 | 8 | 30 | 5 | 180 |
| 609310 | 0 | 72 | 0 | 18 | 76 | 14 | 180 |
| 609720 | 0 | 137 | 0 | 12 | 30 | 1 | 180 |
| 609790 | 0 | 40 | 0 | 19 | 59 | 62 | 180 |
| 609830 | 0 | 43 | 0 | 13 | 94 | 30 | 180 |
| 620160 | 0 | 22 | 0 | 19 | 88 | 51 | 180 |
| 620520 | 0 | 65 | 0 | 22 | 59 | 34 | 180 |
| 630260 | 0 | 64 | 0 | 16 | 81 | 19 | 180 |
| 631000 | 0 | 56 | 0 | 9 | 92 | 23 | 180 |
| 631740 | 0 | 32 | 0 | 10 | 56 | 82 | 180 |
| 631960 | 0 | 38 | $\bigcirc$ | 19 | 90 | 33 | 180 |
| 632220 | 0 | 98 | 0 | 15 | 64 | 3 | 180 |
| 632390 | 0 | 119 | 0 | 15 | 38 | 8 | 180 |
| 642980 | 1 | 49 | 0 | 20 | 84 | 26 | 180 |
| 643490 | 0 | 62 | 0 | 11 | 60 | 47 | 180 |
| 644110 | 0 | 54 | 0 | 22 | 82 | 22 | 180 |
| 644320 | 0 | 49 | $\bigcirc$ | 8 | 83 | 40 | 180 |
| 644480 | 0 | 51 | 0 | 11 | 101 | 17 | 180 |
| 644640 | 0 | 53 | 0 | 18 | 87 | 22 | 180 |
| 654990 | 0 | 70 | 0 | 17 | 78 | 15 | 180 |
| 655570 | 0 | 70 | 0 | 8 | 59 | 43 | 180 |
| 655580 | 0 | 54 | 0 | 6 | 73 | 47 | 180 |
| 655960 | 0 | 47 | 0 | 12 | 85 | 36 | 180 |
| 655970 | 0 | 39 | 0 | 15 | 61 | 65 | 180 |
| 666980 | 0 | 124 | $\bigcirc$ | 8 | 44 | 4 | 180 |
| 667090 | 0 | 50 | 0 | 10 | 75 | 45 | 180 |
| 667370 | 0 | 63 | 0 | 21 | 71 | 25 | 180 |
| 667390 | 0 | 51 | 0 | 8 | 78 | 43 | 180 |
| 667420 | 0 | 48 | 0 | 16 | 94 | 22 | 180 |
| 667530 | 0 | 54 | 0 | 21 | 69 | 36 | 180 |
| 667660 | 0 | 43 | 0 | 7 | 99 | 31 | 180 |
| 677930 | 0 | 72 | 0 | 3 | 87 | 18 | 180 |
| 677940 | 0 | 65 | 0 | 27 | 87 | 1 | 180 |
| 677950 | 0 | 49 | 0 | 9 | 91 | 31 | 180 |
| 678130 | 0 | 43 | 0 | 11 | 63 | 63 | 180 |
| 678140 | 0 | 73 | 0 | 19 | 71 | 17 | 180 |
| 678680 | 0 | 75 | 0 | 11 | 85 | 9 | 180 |
| 678710 | 0 | 96 | $\bigcirc$ | 9 | 56 | 19 | 180 |
| 678940 | 0 | 41 | 0 | 7 | 30 | 102 | 180 |
| 679570 | 0 | 36 | 0 | 10 | 101 | 33 | 180 |
| 901770 | 0 | 71 | $\bigcirc$ | 15 | 63 | 31 | 180 |
| 903170 | $\bigcirc$ | 52 | 0 | 22 | 57 | 49 | 180 |
| 915550 | 0 | 54 | 0 | 15 | 90 | 21 | 180 |
| Nobservd | 1 | 5423 | 1 | 1150 | 6136 | 2589 | 15300 |

Table 8. Rearranged category scores for Festuca.

|  |  |  |  |  |  |
| ---: | :--- | :--- | ---: | ---: | ---: |
|  |  |  | Category |  |  |
|  |  |  | 1 | 2 | 3 |
| 1 | 407030 | PAULINO | 55 | 84 | 41 |
| 2 | 411190 | MUSTANG | 52 | 87 | 41 |
| 3 | 416520 | FA 402 | 79 | 89 | 12 |
| 4 | 416530 | BONAPARTE | 61 | 93 | 26 |
| 5 | 416800 | ISS-0 | 56 | 86 | 38 |
| 6 | 419770 | BAR FA 209 | 77 | 91 | 12 |
| 7 | 423710 | BAR RZ 315 | 54 | 82 | 44 |
| 8 | 423810 | Hykor | 86 | 85 | 9 |
| 9 | 426950 | Kora | 141 | 35 | 4 |
| 10 | 426960 | G 48 | 113 | 56 | 11 |
| 11 | 512670 | KASBA | 102 | 69 | 9 |
| 12 | 525950 | PASTELLE | 88 | 86 | 6 |
| 13 | 526540 | CLARINE | 130 | 43 | 7 |
| 14 | 533760 | BARCEL | 113 | 63 | 4 |
| 15 | 533940 | LUBRETTE | 117 | 58 | 5 |
| 16 | 534010 | ONDINE | 47 | 103 | 30 |
| 17 | 538910 | BARTES | 35 | 87 | 58 |
| 18 | 539130 | SOPLINE | 101 | 69 | 10 |
| 19 | 539310 | DOVEY | 20 | 101 | 59 |
| 20 | 548500 | BUFFALO | 83 | 91 | 6 |
| 21 | 552610 | OLGA | 47 | 93 | 40 |
| 22 | 553110 | CIGALE | 73 | 102 | 5 |
| 23 | 559300 | FUEGO | 54 | 103 | 23 |
| 24 | 559340 | FESTORINA | 36 | 113 | 31 |
| 25 | 559780 | RIVIERA | 104 | 70 | 6 |
| 26 | 560280 | HOUNDOG | 33 | 97 | 50 |
| 27 | 566010 | ADVENTURE | 51 | 101 | 28 |
| 28 | 566270 | FLORINE | 46 | 117 | 17 |
| 29 | 566440 | ARIANE | 92 | 72 | 16 |
| 30 | 572470 | NUBA | 43 | 88 | 49 |
| 31 | 572550 | APACHE | 45 | 113 | 22 |
| 32 | 572590 | FATIMA | 48 | 94 | 38 |
| 33 | 580680 | CLEMFINE | 34 | 100 | 46 |
| 34 | 580770 | DARCY | 85 | 78 | 17 |
| 35 | 581530 | NOVO | 84 | 40 | 56 |
| 36 | 589890 | JAGUAR | 33 | 106 | 41 |
| 37 | 589970 | NORIA | 13 | 37 | 130 |
| 38 | 590060 | SINFONIA | 54 | 96 | 30 |
| 39 | 598740 | SEINE | 13 | 109 | 58 |
| 40 | 599220 | AMELIE | 51 | 104 | 25 |
|  |  |  |  |  |  |


| Table 8. Continued. |  |  |  |  |  |
| :---: | :---: | :--- | ---: | ---: | ---: |
| 41 | 599590 | VILLAGEOIS | 44 | 89 | 47 |
| 42 | 609060 | FELINE | 49 | 94 | 37 |
| 43 | 609300 | GARDIAN | 137 | 38 | 5 |
| 44 | 609310 | IBIS | 72 | 94 | 14 |
| 45 | 609720 | LUTINE | 137 | 42 | 1 |
| 46 | 609790 | MAX | 40 | 78 | 62 |
| 47 | 609830 | WRANGLER | 43 | 107 | 30 |
| 48 | 620160 | MADRA | 22 | 107 | 51 |
| 49 | 620520 | MIRO | 65 | 81 | 34 |
| 50 | 630260 | SILVERADO | 64 | 97 | 19 |
| 51 | 631000 | MURRAY | 56 | 101 | 23 |
| 52 | 631740 | ELDORADO | 32 | 66 | 82 |
| 53 | 631960 | DYNOS | 38 | 109 | 33 |
| 54 | 632220 | MYLENA | 98 | 79 | 3 |
| 55 | 632390 | LUNIBELLE | 119 | 53 | 8 |
| 56 | 642980 | BORNEO | 50 | 104 | 26 |
| 57 | 643490 | ASTERIX | 62 | 71 | 47 |
| 58 | 644110 | COCHISE | 54 | 104 | 22 |
| 59 | 644320 | CARMINE | 49 | 91 | 40 |
| 60 | 644480 | EMPEROR | 51 | 112 | 17 |
| 61 | 644640 | TOMAHAWK | 53 | 105 | 22 |
| 62 | 654990 | ELFINA | 70 | 95 | 15 |
| 63 | 655570 | BARFELIX | 70 | 67 | 43 |
| 64 | 655580 | BARBIZON | 54 | 79 | 47 |
| 65 | 655960 | LEPRECHAUN | 47 | 97 | 36 |
| 66 | 655970 | SAVOY | 39 | 76 | 65 |
| 67 | 666980 | NOBEL | 124 | 52 | 4 |
| 68 | 667090 | PST-RDG | 50 | 85 | 45 |
| 69 | 667370 | VEGAS | 63 | 92 | 25 |
| 70 | 667390 | BARLEDUC | 51 | 86 | 43 |
| 71 | 667420 | BARDOUX | 48 | 110 | 22 |
| 72 | 667530 | DP PL 7901 | 54 | 90 | 36 |
| 73 | 667660 | ZPS J3 | 43 | 106 | 31 |
| 74 | 677930 | BAR RZ 480 | 72 | 90 | 18 |
| 75 | 677940 | BAR FA 411 | 65 | 114 | 1 |
| 76 | 677950 | DTF | 49 | 100 | 31 |
| 77 | 678130 | SG P145 | 43 | 74 | 63 |
| 78 | 678140 | SG P146 | 73 | 90 | 17 |
| 79 | 678680 | FE P174 | 75 | 96 | 9 |
| 80 | 678710 | FE GP30 | 96 | 65 | 19 |
| 81 | 678940 | L.G.M. | 41 | 37 | 102 |
| 82 | 679570 | SFL | 36 | 111 | 33 |
| 83 | 901770 | RABA | 71 | 78 | 31 |
| 84 | 903170 | MANADE | 52 | 79 | 49 |
| 85 | 915550 | LUDION | 54 | 105 | 21 |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
|  |  |  |  |  |  |



Table 10. Numbers of Festuca varieties that were distinct from a particular variety at $\mathrm{p}<0.001$ in an analysis of variance on the cumulative logits for the categories 1 and 2 (AOV1 and AOV2), the coincidence between both analyses of variance (AOV1-2), the numbers found distinct at $\mathrm{p}<0.01$ by a threshold model (TM), and the coincidence between consistent analyses of variance differences (AOV1-2) and threshold model differences (AOV-TM). (* Total is sum of column divided by 2 ).

|  | Number | AOV1 | AOV2 | AOV1-2 | TM | AOV-TM |
| ---: | ---: | :---: | :---: | :---: | ---: | :---: |
| 1 | 407030 | 7 | 17 | 6 | 11 | 6 |
| 2 | 411190 | 7 | 17 | 6 | 12 | 6 |
| 3 | 416520 | 7 | 20 | 5 | 10 | 5 |
| 4 | 416530 | 10 | 5 | 2 | 8 | 2 |
| 5 | 416800 | 13 | 14 | 9 | 11 | 8 |
| 6 | 419770 | 7 | 14 | 5 | 9 | 5 |
| 7 | 423710 | 7 | 20 | 6 | 12 | 6 |
| 8 | 423810 | 6 | 20 | 6 | 12 | 6 |
| 9 | 426950 | 56 | 48 | 46 | 62 | 46 |
| 10 | 426960 | 32 | 13 | 11 | 37 | 11 |
| 11 | 512670 | 19 | 26 | 12 | 23 | 11 |
| 12 | 525950 | 7 | 36 | 7 | 14 | 6 |
| 13 | 526540 | 57 | 39 | 38 | 57 | 38 |
| 14 | 533760 | 32 | 48 | 27 | 42 | 25 |
| 15 | 533940 | 34 | 44 | 27 | 45 | 27 |
| 16 | 534010 | 8 | 14 | 6 | 11 | 6 |
| 17 | 538910 | 14 | 26 | 14 | 24 | 14 |
| 18 | 539130 | 15 | 14 | 8 | 21 | 8 |
| 19 | 539310 | 30 | 25 | 24 | 29 | 24 |
| 20 | 548500 | 7 | 40 | 6 | 12 | 6 |
| 21 | 552610 | 8 | 17 | 7 | 12 | 7 |
| 22 | 553110 | 6 | 44 | 4 | 9 | 4 |
| 23 | 559300 | 7 | 5 | 2 | 9 | 2 |
| 24 | 559340 | 14 | 13 | 8 | 14 | 8 |
| 25 | 559780 | 19 | 40 | 18 | 29 | 16 |
| 26 | 560280 | 13 | 23 | 12 | 20 | 12 |
| 27 | 566010 | 8 | 11 | 5 | 10 | 5 |
| 28 | 566270 | 10 | 6 | 2 | 9 | 2 |
| 29 | 566440 | 10 | 5 | 2 | 12 | 2 |
| 30 | 572470 | 8 | 22 | 8 | 16 | 8 |
| 31 | 572550 | 10 | 6 | 2 | 10 | 2 |
| 32 | 572590 | 15 | 17 | 11 | 12 | 9 |
| 33 | 580680 | 20 | 20 | 14 | 19 | 14 |
| 34 | 580770 | 7 | 12 | 4 | 11 | 4 |
| 35 | 581530 | 8 | 26 | 3 | 8 | 3 |
| 36 | 589890 | 14 | 14 | 9 | 16 | 9 |
| 37 | 589970 | 71 | 78 | 69 | 82 | 69 |
| 38 | 590060 | 7 | 14 | 6 | 10 | 6 |
| 39 | 598740 | 52 | 26 | 25 | 30 | 25 |
| 40 | 599220 | 7 | 9 | 5 | 9 | 5 |
|  |  |  |  |  |  |  |


| 41 | 599590 | 17 |
| ---: | ---: | ---: |
| 42 | 609060 | 8 |
| 43 | 609300 | 66 |
| 44 | 609310 | 6 |
| 45 | 609720 | 67 |
| 46 | 609790 | 10 |
| 47 | 609830 | 10 |
| 48 | 620160 | 29 |
| 49 | 620520 | 7 |
| 50 | 630260 | 7 |
| 51 | 631000 | 13 |
| 52 | 631740 | 22 |
| 53 | 631960 | 10 |
| 54 | 632220 | 14 |
| 55 | 632390 | 41 |
| 56 | 642980 | 10 |
| 57 | 643490 | 7 |
| 58 | 644110 | 10 |
| 59 | 644320 | 8 |
| 60 | 644480 | 10 |
| 61 | 644640 | 10 |
| 62 | 654990 | 7 |
| 63 | 655570 | 7 |
| 64 | 655580 | 10 |
| 65 | 655960 | 13 |
| 66 | 655970 | 14 |
| 67 | 666980 | 55 |
| 68 | 667090 | 8 |
| 69 | 667370 | 7 |
| 70 | 667390 | 9 |
| 71 | 667420 | 8 |
| 72 | 667530 | 7 |
| 73 | 667660 | 15 |
| 74 | 677930 | 6 |
| 75 | 677940 | 7 |
| 76 | 677950 | 10 |
| 77 | 678130 | 14 |
| 78 | 678140 | 6 |
| 79 | 678680 | 6 |
| 80 | 678710 | 19 |
| 81 | 678940 | 10 |
| 82 | 679570 | 15 |
| 83 | 901770 | 6 |
| 84 | 903170 | 7 |
| 85 | 915550 | 7 |
|  |  |  |
|  | Total* | 672 |



Table 11. Modified joint regressions of estimated Variety by Year means in threshold model on Variety main effects.

Response variate: x95

|  | d.f. | s.s. | m.s. | v.r. |
| :--- | ---: | ---: | ---: | ---: |
| Regression | 1 | 40.11 | 40.1094 | 279.03 |
| Residual | 83 | 11.93 | 0.1437 |  |
| Total | 84 | 52.04 | 0.6195 |  |

Percentage variance accounted for 76.8

|  | estimate | s.e. | t (83) |
| :--- | ---: | ---: | ---: |
| Constant | 0.7321 | 0.0505 | 14.50 |
| main | 0.7277 | 0.0436 | 16.70 |

Response variate: x96

|  | d.f. | s.s. | m.s. | v.r. |
| :--- | ---: | ---: | ---: | ---: |
| Regression | 1 | 136.27 | 136.2693 | 441.22 |
| Residual | 83 | 25.63 | 0.3088 |  |
| Total | 84 | 161.90 | 1.9274 |  |

Percentage variance accounted for 84.0

|  | estimate | s.e. | t(83) |
| :--- | ---: | ---: | ---: |
| Constant | -1.3115 | 0.0740 | -17.72 |
| main | 1.3413 | 0.0639 | 21.01 |

Response variate: x97

|  | d.f. | s.s. | m.s. | v.r. |
| :--- | ---: | ---: | ---: | ---: |
| Regression | 1 | 65.64 | 65.6414 | 398.58 |
| Residual | 83 | 13.67 | 0.1647 |  |
| Total | 84 | 79.31 | 0.9442 |  |

Percentage variance accounted for 82.6

|  | estimate | s.e. | t (83) |
| :--- | ---: | ---: | ---: |
| Constant | 0.5795 | 0.0541 | 10.72 |
| main | 0.9310 | 0.0466 | 19.96 |

Figure 1. Cumulative lo for category of category 2 against those category 1.


## cumlogit[1]



Figure 2a. Half-Normal plot for residuals from analysis of variance for cumulative logits of category 1.


Figure 2b. Half-Normal plot for residuals from analysis of variance for cumulative logits of category 2.


Figure 3a. Variety means for 1995 against variety means over full period 1995-1997.


Figure 3b. Variety means for 1996 against variety means over full period 1995-1997.


Figure 3c. Variety for 1997 against variety means over full period 1995-1997.

