

TWC/15/14 Rev. ORIGINAL: English DATE: June 5, 1997

INTERNATIONAL UNION FOR THE PROTECTION OF NEW VARIETIES OF PLANTS GENEVA

# TECHNICAL WORKING PARTY ON AUTOMATION AND COMPUTER PROGRAMS

Fifteenth Session Budapest, June 3 to 5, 1997

REVISED DOCUMENT ON ANALYZING VISUALLY OBSERVED DATA IN TWO GRASS SPECIES

Document prepared by experts from the Netherlands

### 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, y<sub>i</sub>, for a treatment (variety) i are the expression of a continuous underlying variable, U<sub>i</sub>, that cannot be observed and that follows a certain distribution like the normal or the logistic. The underlying variable U<sub>i</sub> 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: 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...C-1), and finally above the last threshold,  $\Box_{i}$ , the observation belongs to the highest category C. The underlying variable U<sub>i</sub> 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  $\log(\gamma_{ic}/(1-\gamma_{ic})) = \theta_c - \sum \beta_i x_{ij}$ , where  $\sum k$  is the probability that the observation lies in one of the categories 1 to c. The distribution of  $\Box_k$  can be obtained from the multinomial. The parameter  $\prod$  stands for the regression coefficient to the explanatory variable  $x_i$  (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,  $\log(\frac{1}{2k}/(1-\frac{1}{2k}))$ , 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  $\Box$  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:  $\log(\gamma_{ic} / (1 - \gamma_{ic})) = \sum \beta_i x_{ii}$  (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,  $\Box_{c}$ , for c=1...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  $\Box_{f}$  and  $\Box_{c}$ . Next, an approximation to a logistic regression is ordinary regression on the logit transformed cumulative category counts,  $\log(y_{ic} / (\Box_{fc} - y_{ic}))$ , which is equivalent to an analysis of variance for our type of data. The residuals of the logistic regressions for  $\Box_{f}$  and  $\Box_{c}$ , 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 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 variance 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 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 p<0.001 are present in the counts of the column AOV1-2. The disctinctnesses of the simultaneous COY-D at p<0.001 were almost completely contained in those of the threshold model at 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 2a and 2b 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  $U_i$ , 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

Sylvain Grégoire is thanked for kindly making available the data on Dactylis and Festuca.

### Reference

McCullagh, P. & J.A. Nelder (1989) Generalized linear models, 2nd. edn., Chapman and Hall.

F.A. van Eeuwijk CPRO-DLO P.O. Box 16 6700 AA Wageningen The Netherlands

Table 1. Obset	rved ca	tegory	scores	for	alter	nativit[]
in Dactylis (C	Cocksfo	ot).				
	2	4	6	7	8	Total
Number						
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	б	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	0	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	б	9	0	17	180
Total	4250	138	286	1	725	5400

Table 2	2. Rearran	ged category s	cores	for
Dactyl	is.			
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
б	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	б
22	678290	MOM DAC 17	167	13
23	678430	87-2	117	63
24	678580	L-DGL 258	150	30
25	678920	К 2 М	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	б
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	<u>⊥</u> 6	20	16
1348090	25	29	25
1371510	8	ΤÜ	./
Total*	193	210	183

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

Variate: Alternativit														
Source of variation d	.f.	s.s.	m.s.	v.r.										
Year.Rep stratum														
Year Residual	2 6	92.1832 7.6822	46.0916 1.2804	36.00 3.27										
year.rep.*Units* stratu	.m													
Cult Cult.Year Residual	29 58 174	466.0963 45.5479 68.1044	16.0723 0.7853 0.3914	41.06 2.01										
Total	269	679.6141												

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

\*\*\* Accumulated analysis of deviance \*\*\*

				mean deviance
	d.f.	deviance	deviance	ratio
Year	2	195.4658	97.7329	65.65
Year.Rep	6	22.0972	3.6829	2.48
Cult	29	1154.2227	39.8008	26.74
Cult.Year	58	113.1846	1.9515	1.95
Cult.Year.Rep	174	191.7480	1.1020	0.75
Residual	5130	3916.1265	0.7634	
Total	5399	5592.8447	1.0359	

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

	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30
Number																														
423320	*	0	0	0	1	0	0	0	1	0	1	0	1	0	0	0	1	0	0	0	0	0	1	0	1	0	0	1	1	0
426850	0	*	-1	-1	0	0	0	0	0	0	0	0	0	0	-1	-1	1	0	0	0	-1	0	1	0	1	0	0	1	1	0
508850	0	1	*	0	1	1	0	0	1	1	1	0	1	1	0	0	1	0	0	1	0	0	1	1	1	1	1	1	1	1
516690	0	1	0	*	1	1	0	0	1	0	1	0	1	1	0	0	1	0	0	0	0	0	1	0	1	1	1	1	1	1
517770	-1	0	-1	-1	*	0	-1	-1	0	-1	0	-1	0	0	-1	-1	1	-1	-1	-1	-1	-1	0	-1	0	0	0	0	1	0
520070	0	0	-1	-1	0	*	0	-1	0	0	0	-1	0	0	-1	-1	1	-1	-1	0	-1	-1	0	0	0	0	0	0	1	0
520080	0	0	0	0	1	0	*	0	1	0	1	0	1	0	0	0	1	0	0	0	0	0	1	0	1	0	0	1	1	0
521780	0	0	0	0	1	1	0	*	1	0	1	0	1	0	0	0	1	0	0	0	0	0	1	0	1	1	0	1	1	0
530200	-1	0	-1	-1	0	0	-1	-1	*	-1	0	-1	0	0	-1	-1	1	-1	-1	0	-1	-1	0	0	0	0	0	0	1	0
559770	0	0	-1	0	1	0	0	0	1	*	0	0	0	0	0	0	1	0	0	0	-1	0	1	0	1	0	0	1	1	0
566170	-1	0	-1	-1	0	0	-1	-1	0	0	*	-1	0	0	-1	-1	1	-1	-1	0	-1	-1	0	0	0	0	0	0	1	0
581380	0	0	0	0	1	1	0	0	1	0	1	*	1	0	0	0	1	0	0	0	0	0	1	0	1	0	0	1	1	0
599260	-1	0	-1	-1	0	0	-1	-1	0	0	0	-1	*	0	-1	-1	1	-1	-1	0	-1	-1	0	0	0	0	0	0	1	0
609050	0	0	-1	-1	0	0	0	0	0	0	0	0	0	*	-1	-1	1	0	0	0	-1	0	1	0	1	0	0	1	1	0
620150	0	1	0	0	1	1	0	0	1	0	1	0	1	1	*	0	1	0	0	0	0	0	1	0	1	1	1	1	1	1
620730	0	1	0	0	1	1	0	0	1	0	1	0	1	1	0	*	1	0	0	0	0	0	1	1	1	1	1	1	1	1
655360	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	*	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1
655650	0	0	0	0	1	1	0	0	1	0	1	0	1	0	0	0	1	*	0	0	0	0	1	0	1	0	0	1	1	0
655870	0	0	0	0	1	1	0	0	1	0	1	0	1	0	0	0	1	0	*	0	0	0	1	0	1	1	1	1	1	0
655880	0	0	-1	0	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	*	-1	0	1	0	1	0	0	1	1	0
667410	0	1	0	0	1	1	0	0	1	1	1	0	1	1	0	0	1	0	0	1	*	0	1	1	1	1	1	1	1	1
678290	0	0	0	0	1	1	0	0	1	0	1	0	1	0	0	0	1	0	0	0	0	*	1	0	1	1	1	1	1	0
678430	-1	-1	-1	-1	0	0	-1	-1	0	-1	0	-1	0	-1	-1	-1	1	-1	-1	-1	-1	-1	*	-1	0	0	0	0	1	-1
678580	0	0	-1	0	1	0	0	0	0	0	0	0	0	0	0	-1	1	0	0	0	-1	0	1	*	1	0	0	1	1	0
678920	-1	-1	-1	-1	0	0	-1	-1	0	-1	0	-1	0	-1	-1	-1	1	-1	-1	-1	-1	-1	0	-1	*	0	0	0	1	-1
903120	0	0	-1	-1	0	0	0	-1	0	0	0	0	0	0	-1	-1	1	0	-1	0	-1	-1	0	0	0	*	0	0	1	0
903150	0	0	-1	-1	0	0	0	0	0	0	0	0	0	0	-1	-1	1	0	-1	0	-1	-1	0	0	0	0	*	0	1	0
1127680	-1	-1	-1	-1	0	0	-1	-1	0	-1	0	-1	0	-1	-1	-1	1	-1	-1	-1	-1	-1	0	-1	0	0	0	*	1	-1
1348090	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	*	-1
1371510	0	0	-1	-1	0	0	0	0	0	0	0	0	0	0	-1	-1	1	0	0	0	-1	0	1	0	1	0	0	1	1	*

	1				~	7.		<u> </u>	
Table	7. Observed	са	tegory	scores	for	altern	atıvıt	∐ın ŀ	estuca.
				Cat	egory				
	_	0	2	3	4	6	8	Total	
	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	0	15	20	4	180	
	426960	0	113	0	9	47	11	180	
	512670	0	102	0	12	57	9	180	
	525950	0	88	0	14	72	6	180	
	526540	0	130	0	6	37	7	180	
	533760	0	113	0	16	47	4	180	
	533940	0	117	0	17	41	5	180	
	534010	0	47	0	19	84	30	180	
	538910	0	35	0	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	2.4	69	40	180	
	553110	0	73	0	10	92		180	
	559300	Ő	54	0	- 0	94	23	180	
	559340	Ő	36	0	11	102	31	180	
	559780	0	104	0	15	55	6	180	
	560280	0	22	0	14	83	50	180	
	566010	0	55	0	10	83	20	180	
	566270	0	46	0	17	100	17	180	
	566440	0	92	0	14	58	16	180	
	500440	0	12	0	12	75	10	100	
	572470	0	43	0	17	06	49	100	
	572550	0	40	0	± /	90	22	100	
	572590	0	40 24	0	14	00	30 46	100	
	500000	0	54	0	14	60	40	100	
	580770	0	85	0	15	03		100	
	581530	0	84	0	14	34	50	100	
	589890	0	33	0	14	92	41	100	
	589970	0	13	0	0	37	130	180	
	590060	0	54	0	14	82	30	180	
	598740	0	13	0	6	103	58	180	
	599220	0	51	0	23	81	25	180	
1									

Table 7. Cont	inued.							
EQQEQO	0	4.4	0	11	70	17	100	
609060	0	19	0	11	83		180	
609300	0	137	0	8	30	5	180	
609310	0	72	0	18	76	14	180	
609720	0	137	Ő	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	0	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	0	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 10	/3	4/	180	
655960	0	4 /	0	1	85 (1	30	100	
655970	0	104	0	0 C T	01	20	100	
667090	0	50	0	0 10	44	4	180	
667370	0	50	0	21	73	J - J	180	
667390	0	51	0	21	71	43 43	180	
667420	0	48	0	16	94	22	180	
667530	0	54	0	21	69	36	180	
667660	0	43	Ő	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	0	9	56	19	180	
678940	0	41	0	7	30	102	180	
679570	0	36	0	10	101	33	180	
901770	0	71	0	15	63	31	180	
903170	0	52	0	22	57	49	180	
915550	0	54	0	15	90	21	180	
Nobservd	1	5423	1	1150	6136	2589	15300	

Table	<b>8.</b> Rea	arranged cat	egory	scores	for Fest	uca.
				Catego	orv	
			1	2	3	
1	407020	DATIT TNO	55	01	11	
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	
10	426950	Kora	141	35	4	
10	420900	G 48 KAGBA	102	50	11	
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 552110	OLGA	4/	102	40	
22	559300	FUEGO	54	102	23	
2.4	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	580680	FAIIMA	48	100	38 46	
33	580770	DARCY	24	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	O Com	tinued				
Table	<b>8.</b> Con	LINUED.		0.0	4.5	
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	DVNOS	38	109	22	
54	632220	MVLENA	98	79	3	
55	622220		110	52	0	
55	642090	DODNEO	TT 2	104	26	
50	642980	DORINEO	50	104 71	20	
57	643490	ASIERIA	02	104	4 /	
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 D145	42	74	63	
7.9	678140	SG D146		90	17	
70	678680	FF D174	75	96	, T	
00	670710		15	90 6 F	9 10	
00	670040	FE GPSU	90 41	20	100	
0 D	0/0940	L.G.M.	4⊥ 2¢	3/ 111	707 707	
8∠	0/95/0	SFL DADA	30	111	33	
83	9017/0	KABA	71	78	31	
84	903T./0	MANADE	52	./9	49	
85	915550	LUDION	54	105	21	
			5424	7287	2589	

Table 9. Rea	irrange	d c	ategor	y sco	ores	for	Festu	ca,	per	year.								
Number	1	95 2	3	1	96 2	3	1	97 2	3									
Number 1 407030 2 411190 3 416520 4 416530 5 416800 6 419770 7 423710 8 423810 9 426950 10 426960 11 512670 12 525950 13 526540 14 533760 15 533940 16 534010 17 538910 18 539130 20 548500 21 552610 22 553110 23 559300 24 559340 25 559780 26 560280 27 566010 28 566270 29 566440 30 572470 31 572550 32 572590 33 580680 34 580770 35 581530 36 589890 37 589970 38 590060 39 598740 40 599220	1 6 11 17 1 15 7 19 41 33 16 12 31 28 34 6 3 24 10 11 8 12 32 8 6 0 17 10 6 11 15 7 19 41 33 16 12 31 28 34 6 3 24 10 11 15 7 19 41 33 16 12 31 28 34 6 3 24 10 11 17 10 11 10 10 11 10 10 11 10 10	2 44439443199535826994608283114455235052024874	3 7 9 0 10 19 2 2 0 2 1 3 1 0 0 5 28 2 13 0 11 0 5 28 2 13 0 11 0 5 28 2 13 0 11 0 5 28 2 13 0 11 0 5 5 28 2 10 10 2 12 2 0 2 13 1 0 0 5 5 28 2 10 10 10 2 12 2 0 2 13 1 0 0 5 5 2 8 2 10 10 10 5 5 2 8 2 10 0 11 0 5 5 2 8 2 10 0 11 0 10 10 10 10 10 10 10 10 10 10	1 37445748824430218441294377794446648271791361130	2 6 14 10 2 8 14 11 5 7 6 7 1 9 8 12 13 12 19 13 17 10 15 14 16 4 0 18 4 11 19 13 17 19 19 13 12 19 13 17 19 13 12 19 13 17 19 13 12 19 13 17 19 13 12 19 13 17 19 13 17 19 19 13 12 19 13 17 19 13 12 13 12 13 12 13 15 14 16 17 19 13 17 19 18 19 13 17 19 13 17 10 15 15 14 16 17 19 15 15 14 16 17 19 18 17 10 15 15 14 16 15 15 14 16 19 15 15 14 16 15 15 14 16 15 15 14 16 15 15 15 14 16 15 15 15 15 15 15 15 15 15 15	3 17 12 5 1 4 8 20 5 2 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 3 12 12 3 12 12 3 12 12 3 12 12 12 12 12 13 12 12 12 13 10 4 25 3 11 14 22 12 3 11 14 22 12 3 11 14 22 12 3 11 14 22 12 3 11 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 22 13 14 23 15 14 15 15 14 15 15 15 15 15 15 15 15 15 15	1 12 77 37 48 24 30 45 44 44 44 44 44 44 44 44 44	2 1336284051404424697454731657496066666666666666666666666666666666666	$\begin{array}{c} 3 \\ 170 \\ 150 \\ 22 \\ 22 \\ 66 \\ 15 \\ 22 \\ 22 \\ 66 \\ 15 \\ 22 \\ 22 \\ 66 \\ 15 \\ 22 \\ 22 \\ 10 \\ 41 \\ 3 \\ 99 \\ 99 \\ 99 \\ 99 \\ 99 \\ 99 \\ 99$	Continue 0 0 0 0 40 0 11 0 33 0 10 0 12 0 12 0 12 0 12 0 12 0 12 0 12	$\begin{array}{c} \mathbf{a} = \mathbf{c} \cdot \mathbf{c} \\ \mathbf{a} = \mathbf{c} \\ $	5   31   32   59   0   49   51   50   46   52   24   51   50   46   29   23   42   51   50   52   27   50   51   50   52   29   51   50   51   50   51   50   51   50   51   50   51   51	9 10 12 22 12 22 12 22 9 7 16 18 8 4 11 5 5 13 9 12 18 1 4 10 9 12 12 22 12 22 9 7 16 18 8 4 11 22 12 22 12 2 9 7 16 18 18 12 12 2 12 2 9 7 16 18 18 1 12 2 18 1 1 2 18 1 1 2 18 1 1 2 18 1 1 2 18 1 1 2 18 1 1 1 2 18 1 1 1 2 18 1 1 1 2 18 1 1 1 1 2 18 1 1 1 2 18 1 1 1 2 18 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1	20 18 0 4 12 12 14 12 14 10 12 14 10 12 14 10 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 12 14 15 15 14 12 14 15 15 14 15 15 15 15 15 15 15 15 15 15	13 12 38 17 45 11 12 5 3 8 28 42 7 10 5 5 7 2 22 16 6 5 2 41 7 8 3 9 10 11 5 2 21 6 5 2 3 17 10 11 5 13 10 12 13 10 12 10 10 11 12 10 10 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 11 10 10	36 37 315 37 397 377 397 377 397 377 397 377 397 377 397 377 397 377 397 315 327 397 315 327 397 315 327 397 315 327 327 327 327 327 327 327 327	$\begin{array}{c}11\\16\\5\\0\\22\\11\\16\\40\\13\\1\\1\\6\\25\\7\\10\\5\\8\\6\\25\\14\\10\\8\\0\\11\\26\\5\\2\\15\\2\\8\\17\\18\\10\end{array}$	

**Table 10.** Numbers of Festuca varieties that were distinct from a particular variety at 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 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	ΤM	AOV-TM	
1	407030	7	17	6	11	б	
2	411190	7	17	б	12	б	
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	б	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	б	11	б	
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	б	
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	б	2	9	2	
29	566440	10	5	2	12	2	
30	572470	8	22	8	16	8	
31	572550	10	б	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	ן ד
50	631000	12
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	667520	87
72	667660	15
74	677930	15
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	б
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 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.



TWC/15/14 Rev.

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.

[End of document]