FACTOR ANALYSIS OF PLANT-TYPE VARIABLES RELATED TO YIELD OF DRY BEANS (PHASEOLUS VULGARIS L.)

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This is to certify that the

thesis entitled

Factor Analysis of Plant-Type Variables Related to Yield of Dry Beans (Phaseolus Vulgaris L.) presented by

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ABSTRACT

FACTOR ANALYSIS OF PLANT-TYPE VARIABLES RELATED TO YIELD OF DRY BEANS (PHASEOLUS VULGARIS L.)

Ву

Jacques Clarel Denis

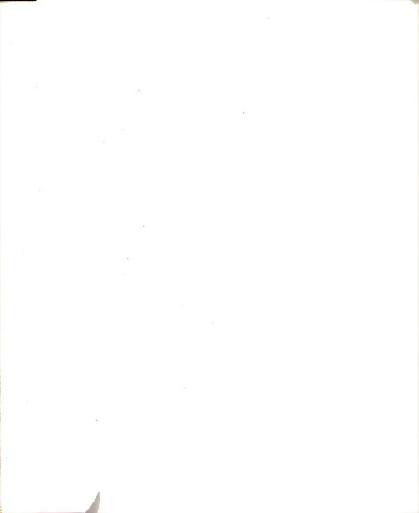
Negative correlations between the components of yield, namely number of pods per plant (X), number of seeds per pod (Y) and seed weight (Z), are often seen as major barriers to yield improvement in beans. A question is raised as to whether these traits are part of a larger set of variables which determine collectively the pattern of production of the bean plant. In trying to answer this question, an image factor analysis was carried out on a set of twenty-two variables involved in fixing the yield potential of the plant.

The material consisted of sixteen varieties, of which eight are determinate, eight indeterminate or six are light-seeded, five have medium seed weight and five are heavy-seeded lines. They were grown at two locations in Michigan during the summer of 1970. A randomized block design with four replications and single-row plots was used at both locations. But, in order to accentuate the differences between the two environments, a standard number of sixty seeds per plot was sown at East Lansing, the first location, whereas a standard weight of fifteen grams was planted at Gratiot, the second location. The rows were always

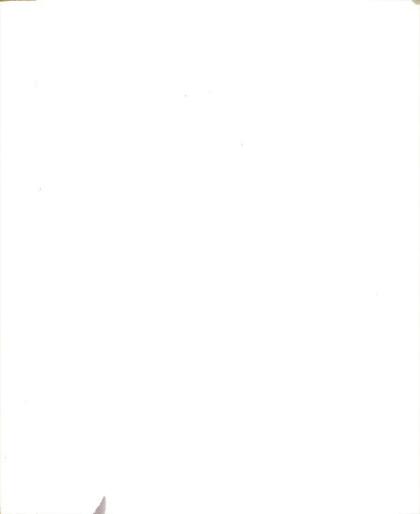
6 meters long and 50 centimeters apart. Observations were made at maturity and all analyses were performed on the logarithms of the averages over five plants chosen at random in every plot. The twenty-two traits obtained from these observations and included in this study were : total number of nodes per plant, number of nodes per plant with pod-bearing branches, total number of branches per plant, number of racemes on the branches per plant, number of pods on the branches per plant, total number of racemes per plant, total number of pods per plant, average number of seeds per pod, average seed weight, average number of pods per productive node, average plant weight, average pod weight, average pod breadth, average pod thickness, average pod length, number of bassal or short internodes, number of upper or long internodes, average short-internode length, average long-internode length, average hypocotyl diameter, average short-internode diameter, average long-internode diameter, numbered in the same order they are listed here. A full model of factor analysis was assumed and squared multiple correlations were used as communalities. The correlation matrix with these communalities as elements of the leading diagonal was then adjusted and became the image covariance matrix, the input to image analysis. The analyses were performed on total variation, and only the factors with eigenroot greater than unity were extracted. The results obtained with the complete-set of data were checked through analyses of sub-sets of data, namely analyses by location, by growth habit, by seed-weight and double-checked through the use of multiple discriminant analysis.

Three major factors or patterns of production were found and together they accounted for 83% of the total variation before rotation and about 77% after rotation. The first two were about evenly important and each extracted after rotation 31% of the variance. The first factor or pattern of production was identified as pod-weight, and characterized a plant-type that will reach a high yield through a high seed weight. This plant has long and sturdy basal internodes, no or very few long internodes and it produces very large pods. The second pattern of production identified a plant-type capable of reaching a high yield through a high number of pods per plant. Like the pod-weight type, the pod-number type has no or very few long internodes but it has instead many moderately sturdy short internodes. Too many nodes will provoke a reduction in the yield of either type. It has been noted that although these two types can be inproved to yield more, far better yield can be obtained with intermediate plant-types resulted from inter-breeding between them. New varieties can thus be produced that will approximate the ideals of the plant breeder for maximum yield on a per-plant basis or an area basis, depending on the number of branches judged acceptable by assuming that the greater their number, stronger a competitor the plant might be. The third factor, a growth factor, characterized by an excessive vegetative growth coupled with a very low production, was seen as a limit to this inter-breeding.

The factors leading to these conclusions are statistically uncorrebut they are not believed to be so from a biological standpoint. So, it



would be very interesting to submit these factors to an oblique rotation and determine their natural association. This can be seen as an important next step, in view of its implication on the possibility of combining maximally the characteristics of the two major plant-types encountered here.



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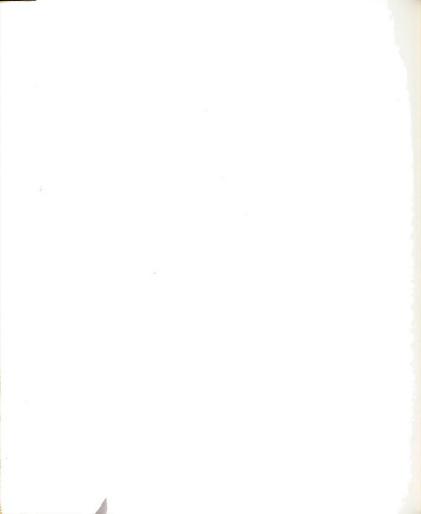
Jacques Clarel Denis

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This thesis is dedicated to those whose labor or thoughts have contributed to the advancement of Science in general, and Agronomy in particular, but whose names, at some point in history, for one reason or another, have ceased to be remembered.



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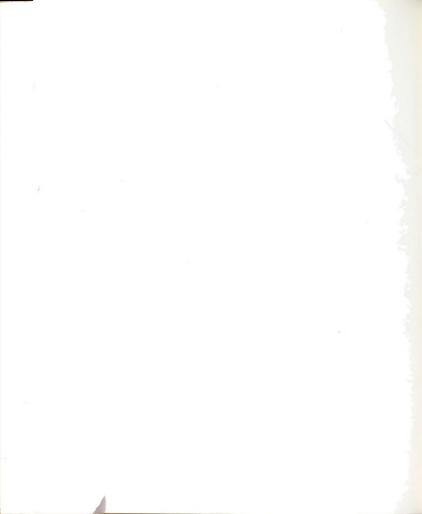
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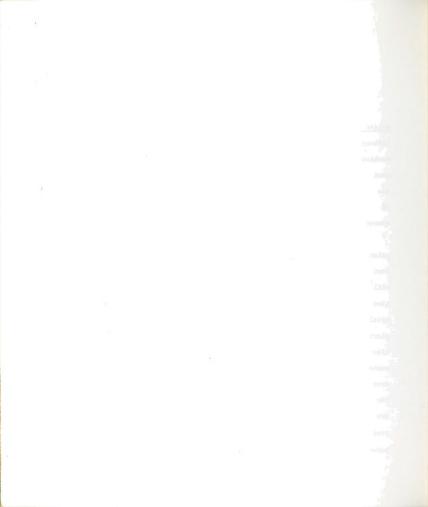


INTRODUCTION

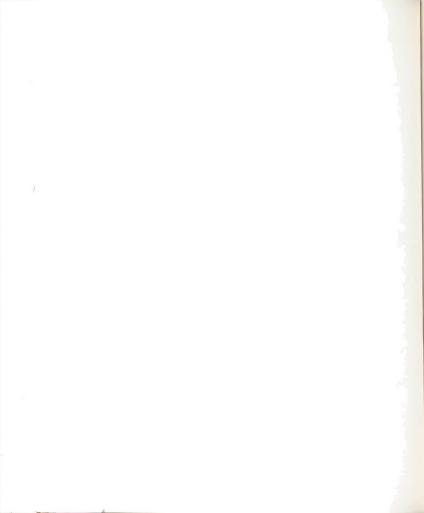
The yield plateau of the present varieties of dry beans, <u>Phaseolus vulgaris L.</u>, is low compared to that of many other field crops.

Although several reasons for low yields in beans can be adduced, I shall mention two problems that are seen as major barriers to yield improvement: one, the physical limitation of the present bean plant-types; and two, the faulty synchionization that often prevails between the demands of the plant for environmental resources and the availability of these resources to the plant over time.

Bean varieties are either bush or vine. Both types can produce a relatively high number of flowers but retain a much reduced number after fertilization. Of these retained and transformed into pods, still fewer will reach the harvest stage. This explains in part the close association generally observed between the number of pods per plant (X) and the total seed yield (W). Seldom can this association be exploited in a plant yield improvement program however, since along with it there exist also negative correlations between the yield components, namely the number of pods per plant (X), the number of seeds per pod (Y) and the average seed weight (Z); here yield is taken as the product of these three components. Some of the questions that can be raised at this point are: Is it the waste of nutrients occasioned in part by the excessive production of flowers which is responsible for the negative correlations between yield components, thus implying some kind of limit on food



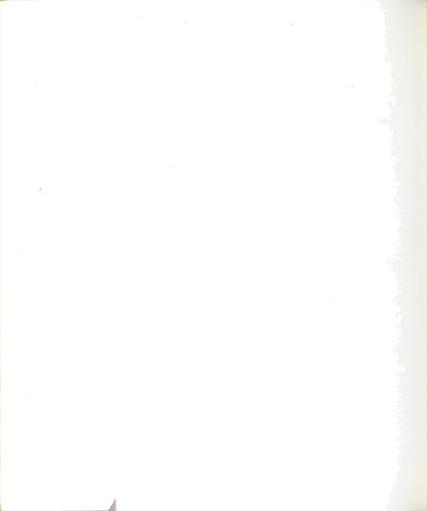
availability together with a marked competition between the components for that food? Or, is it instead a differential through time in the amount and quality of environmental resources available and/or in the photosynthetic capacity of the plant which is the main reason for both the high correlation between X and W and the negative correlations between the components? These two questions are relevant particularly to the second problem mentioned above. But, one can also ask: Are these components the only variables involved or are they only part of a much larger set of traits which together determine the pattern and level of production of the bean plant. To answer this question is to determine whether these patterns exist or not and if so, how they are associated with the different bean plant types. This is the main objective of the present work and it is seen as a first step in the understanding of the problem of physical limitations of the present bean plant types, referred to above. Factor analysis is the statistical tool that will be used to this end.



REVIEW OF LITERATURE

Factor analysis is a statistical technique which facilitates the interpretation of the interrelationships among many variables (Harman, 1968). It consists in the reduction of a large number of observed correlated variables to a much smaller number of hypothetical uncorrelated variables called factors. Scientifically, these factors or constructs are the basic influences or dimensions in the data which can be used later instead of the more numerous original variables since they explain most of the correlations among them (Cattell, 1965).

There exist many methods of factor analysis. They will differ depending on the assumptions made or the model chosen and the specific goal being pursued. However, on the basis of their order of execution, they can be divided between direct solutions methods and derived solutions methods (Harman, 1967). The direct solutions methods would be those used to transform a correlation matrix into a factor matrix, i.e., to find the common-factor space of that correlation matrix. The derived solutions methods concern a reorientation or rotation of the reference axes of that common-factor space, so as to facilitate the interpretation of the results, hence to approach a "simple structure" representation. In other words, the direct or unrotated solution defines the pattern of variance in the sample data, whereas the derived or rotated solution reveals the clusters of intercorrelations among the variables (Rummel, 1970). In the present work, a version of principal-factor analysis, image analysis, and varimax orthogonal rotation have been used as direct



solutions and derived solutions methods, respectively.

Principal-factor analysis as performed today on electronic computers has been developed by Kelley (1935) from the method of principal components of Hotelling presented in 1933. The first computer application was made by Wrigley and Neuhaus in 1952. The original method itself has its foundation in the "method of principal axes" of Karl Pearson (1901), where he showed how a large set of data can be reduced empirically so that a maximum portion of the variance present in the data can be accounted for. Hotelling, in turn, made of this technique the principal components method. However, the father of factor analysis is considered to be Charles Spearman who dedicated forty years of his life to its development (Harman, 1967, 1968). Principal components analysis, as mentioned previously, is a method by which many correlated variables which together can possess some degree of redundancy are reduced to a few uncorrelated variables (the principal components), thus eliminating that redundancy (Bailey, 1956). There will be as many principal components or factors as there are variables in the analysis, although only a few are needed to account for a large part (90% or more) of the sample variance. All of these components should be included, however, in order to reproduce the correlations among the variables. According to Maxwell and Rao, cited individually by Murty and Arunachalam (1967), no assumption is made concerning the distribution of the variables or their randomness. In principal-factor analysis, however, it is assumed that the observed variables follow a multi-variate normal distribution and that the newly established factors (fewer in number than the variables) are linear and additive.

The principal components analysis model is:



$$z_{j} = a_{j1} F_{1} + a_{j2} F_{2} + \cdots + a_{jn} F_{n}$$
 (j = 1, 2, ...,n)

where the observed variable Z , is expressed as a linear regression on \underline{n} j factors (F's). The factors are considered as random variables and they can be defined by a normal probability density function (Harman, 1967). Harman referred to this model as a "maximum variance model". By contrast, the classical factor analysis model is constructed so as to reproduce most of the correlations among the variables. It can be written as follows:

$$z_{j} = a_{j1} F_{1} + a_{j2} F_{2} + \cdots + a_{jm} F_{m} + d_{j} U_{j}$$
 (j = 1, 2, ...,m),

where each observed variable is described as a linear regression on \underline{m} factors only (\underline{m} smaller than \underline{n}) plus a unique factor ($U_{\underline{j}}$). The common factors result from the correlations among the variables whereas the unique factor represents the remaining variance (including error) of a particular variable not accounted for in the set of correlations. A good distinction between the implications of the two models is explicit in this characterization by Cattell, who called the principal component model a closed model and the principal-factor one an open model. In practice however, the method of "principal factors" is known as an application of principal components analysis to a reduced correlation matrix, that is a matrix which has communalities in its main diagonal instead of unities (Cattell, 1965; Harman, 1967). The rest of the correlation matrix is further adjusted in image analysis (Veldman, 1967, Rummel, 1970).

The communalities can be estimated in many ways. Rummel discussed at some length three categories of them. He reported also the squared multiple correlation as being the most commonly used communality estimate (Rummel, 1970). As noted by Harman, Thomson (1934) was the first to use Hotelling's principal components method with the full model

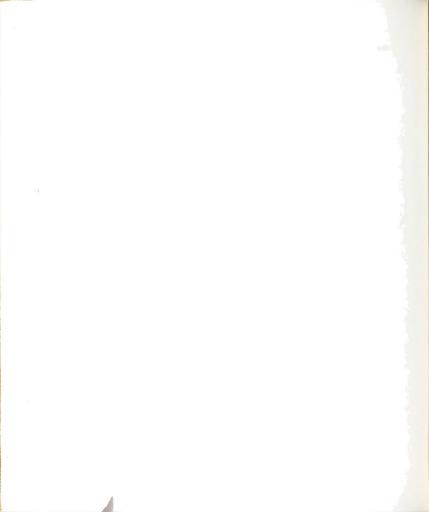


of classical factor analysis (Harman, 1967)

Factor analysis has a long list of very diverse fields of application which includes international relations, sociology, economics, communications, taxonomy, biology, geology, meteorology and particularly psychology and the behavioral sciences (Rummel, 1970, Harman, 1967, 1968). However, most of these applications are as recent as the availability of compouter programs. In the last four or five years, a few works related to plant breeding have been published. For example, in 1967, Murty and Arunachalam, using factor analysis found that the pattern of divergence in the genus Sorghum differs under natural and artificial selection (Murty and Arunachalam, 1967). Later, joined by Jain, they established that, this was the case for both cross and self-pollinated They also determined that selection, be it natural or artificial, has more influence on the factor loadings than the breeding system. Of greater importance was their finding that for both outbreeders and inbreeders, under natural or artificial selection, the variables associated with productivity were the ones with highest loadings on the first factor (Murty, Arunachalam and Jain, 1970). Morishima, Oka and Chang, working with samples from an $F_{\rm q}$ population of a cross between a tall and a short strain of rice, applied principal component analysis to a matrix of average genetic correlations between seven morphological traits. After rotation to "simple structure", they identified two bipolar axes of variation in rice plant type, namely the panicle-number versus paniclelength and the internode-length versus internode-number axes. panicle-number and the internode-length types had higher yielding potential (Morishima, Oka and Chang, 1969). Using the same material, Hashigushi and Morishima performed a principal components analysis on



the phenotypic correlations. They used their own method to determine from the results what they called "the genetic vectors". Interestingly, and significantly, these were comparable to the components extracted from the genetic correlations (Hashigushi and Morishima, 1969). Hegman and DeFries found different results with phenotypically standardized phenotypic, genetic and environmental covariances than with the unstandardized covariances (Hegman and DeFries, 1970). This shows clearly that scaling can have an effect on results from factor analysis. The differences in size of loadings in the principal components extracted from genetic and environmental correlations, in the work by Murty and Arunachalam, seem to indicate the importance of normality as a requirement in factor analysis (ibid.). Consequently, these authors express the view that the environmental correlations are more appropriate for factor analysis.



MATERIALS AND METHODS

Twenty-four homozygous lines of dry edible beans were planted at two locations in Michigan during the summer of 1970. Single-row plots in a randomized block design with four replications were used. The rows were 6 meters long and 50 centimeters apart. In order to accentuate the differences between the two environments, a standard number of seeds (sixty per plot) was sown at East Lansing, the first location, and a standard weight of seeds (fifteen grams per plot) was planted at Gratiot, the second location. I believe that the light-seeded lines are usually favored by the first, more popular type of planting, so I provoked deliberately the opposite situation in the second type of planting. Observations were to be recorded at maturity, because it is the proper time to have a measurement of some of the traits in which I was interested. As an indication of maturity, it was taken that most of the plants in the plot would have to have almost all the pods fully developed and, in addition, 25 to 50% of them should be at the pale green or yellowish color stage. Because of frost damage to different varieties at the two locations, complete sets of data have been obtained on a smaller number of entries than anticipated. Consequently, the present work includes sixteen varieties for which all data have been recorded at both locations. These varieties are listed below. Eight are bush (B) or determinate and eight are vine (V) or indeterminate, or six are



light-seeded (L), five have medium seed weight (M) and five are heavy-seeded (H) lines. Incidentally, the light-seeded varieties correspond to navy beans, the medium seed weight to great northern beans and the heavy-seeded lines to kidney beans.

The sixteen varieties were:

1.	02-	Great Northern 1-27	(M, V)
2.	03 –	Algarrobo	(H, B)
3.	04-	Sanilac	(L, B)
4.	05-	Charlevois	(H, B)
5.	08-	Criolla	(L, B)
6.	09-	Dominican	(L, V)
7.	10-	Mich. Cranberry	(H, V)
8.	12-	Red Mexican	(M, V)
9.	13-	Michelite-62	(L, V)
10.	14-	Perry Marrow	(M, V)
11.	15-	Seafarer	(L, B)
12.	16-	Pinto 114	(M, V)
13.	17-	Manitou (RKBO255)	(H, B)
14.	18-	Rinson Oscuro	(M, B)
15.	21-	Saginaw	(L, V)
16.	23-	Estrada Rosada	(H, B)

At maturity, five plants per line were chosen at random and observations were made on each plant on a node per node basis. However, all analyses were performed on logarithms of the averages over the 5 plants because of the disparity in the variances of the different variables.



Twenty-two traits were finally retained after elimination of complex traits like total seed yield and number of seeds per plant which would probably make the results less interpretable. Also, some traits were not used in order to prevent singularity or too low communality in the final set of traits. The twenty-two traits or variables together determine yield, the coefficient of determination exceeded 98%. These variables were:

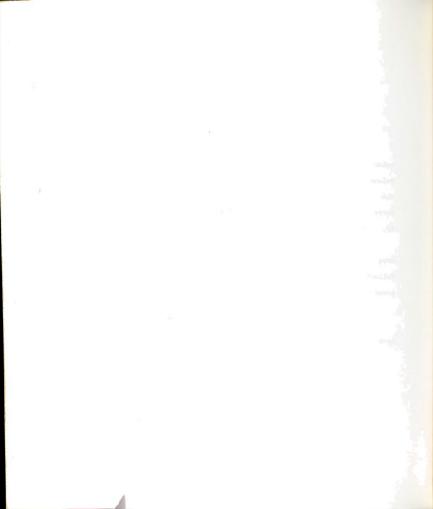
- 1. total number of nodes per plant
- 2. number of nodes with pods on branches per plant
- 3. total number of branches per plant
- 4. number of racemes on the branches per plant
- 5. number of pods on the branches per plant
- 6. total number of racemes per plant
- 7. total number of pods per plant (X)
- 8. average number of seeds per pod (Y)
- 9. average seed weight (Z)
- 10. average number of pods per productive node
- 11. plant fresh weight
- 12. average pod fresh weight
- 13. average pod breadth
- 14. average pod thickness
- 15. average pod length
- 16. number of short-internodes*

Those immediately above the hypocotyl (5 or 6 on the average), which are followed in turn by the long-internodes, more variable in number.



- 17. number of long-internodes
- 18. average short-internode length
- 19. average long-internode length
- 20. average hypocotyl diameter
- 21. average short-internode diameter
- 22. average long-internode diameter

Principal-factor analysis comprises two main parts, namely the calculation of the correlation matrix (R) and the extraction of the eigenvalues and eigenvectors of that correlation matrix. In factor analysis, the use of the correlation matrix as calculated implies that all the correlations result from common factors or common variance. Consequently, neither specific factor of a variable nor error is accounted for. Such an assumption has been discarded in this work. So, the image covariance matrix was used instead of the correlation matrix, thus following the full model of classical factor analysis. The image covariance matrix is a transformed correlation matrix in which the unities in the main diagonal have been replaced by the communalities of the variables, here their squared multiple correlations, and then the rest of the matrix is adjusted accordingly. Then the Jacobi method was applied to the image covariance matrix to extract its eigenroots and eigenvectors. The eigenroots or eigenvalues are the variances or portions of the total variation extracted by the corresponding eigenvectors or principal components. Not all the components can be considered as important, since the first ones are always associated with a greater part of the total variation. So, an eigenroot of at least one was chosen as requirement for an eigenvector to be extracted. After



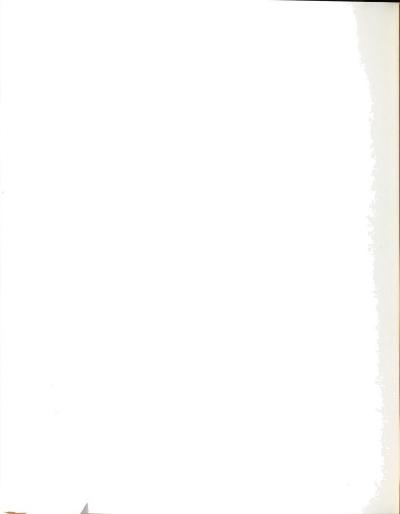
extraction, the factors were submitted to varimax orthogonal rotation.

Varimax as opposed to quartimax rotation simplifies the factor instead of the variables, and this is done so that the orthogonality between any two factor-vectors is preserved.

All analyses were made using the total variation in the sample and assuming the full model of factor analysis. The discussions of the results will concern only the common part of that variation. the complete set of data, analyses were also performed on sub-sets of the data, namely, each location separately, each growth type separately, and finally each seed-weight group separately. The factors extracted for groups within each classification were then compared. The varimax factor loadings were used as bases for these comparisons. of results were obtained here, the correlations between the factorvariables in the two sets of factors being compared, and the correlations between all pairs of variables in the two sets (same variables in both). The diagonal elements of this last matrix are correlations of a given variable in a particular sub-set with the same variable in the other sub-set. Prior to the calculation of these correlations the second structure or set of factors is generally transformed to attain maximum overlap with the first. Finally the results of these comparisons are double-checked through the use of multiple discriminant ana-Since this analysis can indicate which variables in a set account for differences between groups of subjects, when these differences exist, it should be possible, through it, to verify the conclusions arrived at after finding and comparing the rotated factors.



The computer programs used in this work are part of a package offered by D. J. Veldman in his book "Fortran Programming for the Behavioral Sciences" (Veldman, 1967).



RESULTS

Factor Analysis of the Complete Set of Data

The results from the complete set of data indicate that 92.70% of the total variation is common, that is, the common factors can account for that much of that variation, and the remaining 7.30% should be attributed to unique factors and errors. The trace was 20.4, 83.16% of which was extracted by only 3 roots. The first two roots accounted for more than 73% before rotation and more than 62% after rotation (Table 1, page 15). So, it can safely be said that there are mainly two factors at work in the present set of data. The varimax rotation results show the third factor increasing in importance (from about 10% before rotation to about 15% after), whereas the first two become about evenly important (31% each).

The variables with the highest loadings in the first factor are: average pod thickness (14)*, average seed weight (9), average pod fresh weight (12), average pod breadth (13); next in importance are: average pod length (15), average short-internode length (18), all three diameter-variables (20-22) and plant fresh weight (11) (Table 1 and Figure 1 on page 15). Number of seeds per pod (8) has a moderately high and negative loading in this factor. The variables with the lowest loadings are: for the positive, number of nodes with pods on the branches (2), number of

 $^{^{\}star}$ The numbers in parentheses identify the variables as listed on pages 10-11.



Table 1. Factor Analysis Results, Complete Set of Data

Unrotated Factors						otated Fac	tors
ts \$ C*	1	2	3	% ℃*	11	2	3
97.61	6237	.3475	.6117	\$8.39	'3250	.2509	. 8458
97.24	.1393	.8606	.1122	77.25	.0364	.8739	.0867
96.26	.2890	.8062	.0306	73.45	.1394	.8437	0563
99.64	.1676	.9337	.1128	91.27	.0499	.9508	.0788
99.49	.1598	.9539	0798	94.19	0473	.9653	0883
99.44	1459	.9241	.1992	91.50	1844	.8906	.2963
99.42	1965	.9184	0736	88.75	3515	.8708	.0756
86.24	3906	.4619	- .3270	47.29	5641	.3830	0891
96.68	.9183	1573	.2317	92.17	.9355	.0036	2158
91.22	.3075	.6357	4277	68.16	0249	.6691	4830
94.51	.6510	.5728	.1286	76.85	.5385	.6763	1455
98.68	.9102	1082	.2403	89.79	.9245	.0508	2016
92.38	.8268	2223	. 3254	83.90	.9080	0738	0948
96.39	.8527	1443	.3651	88.12	9363	.0082	0565
91.60	.8407	0659	0157	71.13	.7408	.0755	3961
76.56	3571	.5201	.0074	39.81	3924	.4530	.1971
97.59	6746	.2596	.6119	89.69	3556	.1557	.8638
80.80	.6273	3283	.2595	56.86	.7200	2131	0699
67.71	3047	0189	.5800	42.96	0014	0575	.6529
95.73	8649	.3333	0586	86.26	.6792	.4722	4223
96.54	.7904	.4386	.0122	81.72	.6292	.5649	3197
87.56	.8219	0134	2952	76.28	.5893	.1182	6337
ots	8.1827	6.7969	1.9783				
ice	40.13	33.33	9.70	% Var.	31.28	31.03	14.78
Trace	40.13	73.46	83.16	Cum & Var.	31.28	62.31	77.09
Trace =							
	97.61 97.24 96.26 99.64 99.49 99.42 86.24 96.68 91.22 94.51 98.68 92.38 96.39 91.60 76.56 97.59 80.80 67.71 95.73 96.54 87.56	97.616237 97.24 .1393 96.26 .2890 99.64 .1676 99.49 .1598 99.441459 99.421965 86.243906 96.68 .9183 91.22 .3075 94.51 .65510 98.68 .9102 92.38 .8268 96.39 .8527 91.60 .8407 76.563571 97.596746 80.80 .6273 67.713047 95.73 .8649 96.54 .7904 87.56 .8219	ts \$ C* 1 2 97.61 6237 .3475 97.24 .1393 .8606 96.26 .2890 .8062 99.64 .1676 .9337 99.49 .1598 .9559 99.44 1459 .9241 99.42 1965 .9184 86.24 3906 .4619 96.68 .9183 1573 91.22 .3075 .6357 94.51 .6510 .5728 98.68 .9102 1082 92.38 .8268 2223 96.39 .8527 1443 91.60 .8407 0659 76.56 3571 .5201 97.59 6746 .2596 80.80 .6273 3283 67.71 3047 0189 95.73 .8649 .3333 96.54 .7904 .4386 87.56 .8219 0134	ts \$ C* 1 2 3 97.61 6237 .3475 .6117 97.24 .1393 .8606 .1122 96.26 .2890 .8062 .0306 99.64 .1676 .9337 .1128 99.49 .1598 .9539 0798 99.44 -1459 .9241 .1992 99.42 -1965 .9184 0736 86.24 -3906 .4619 -3270 96.68 .9183 1573 .2317 91.22 .3075 .6357 4277 94.51 .6510 .5728 .1286 92.38 .8268 2223 .3254 96.39 .8527 1443 .3651 91.60 .8407 0659 0157 76.56 3571 .5201 .0074 97.59 6746 .2596 .6119 80.80 .6273 3283 .2595 67.71<	\$ C* 1 2 3 \$C* 97.61 6237 .3475 .6117 \$8.39 97.24 .1393 .8606 .1122 .77.25 96.26 .2890 .8062 .0306 .73.45 99.64 .1676 .9537 .1128 .91.27 99.49 .1598 .9559 0795 .94.19 99.44 -1459 .9241 .1992 .91.50 99.42 -1965 .9184 0736 .88.75 86.24 3906 .4619 3270 .47.29 96.68 .9183 1573 .2317 .92.17 91.22 .3075 .6357 4277 .68.16 94.51 .6510 .5728 .1286 .76.85 98.68 .9102 -1082 .2403 .89.79 92.38 .8268 -2223 .3254 83.90 96.39 .8527 -1443 .3651 88.12 91.60	\$ C* 1 2 3 \$ C* 1 97.61 6237 .3475 .6117 \$8.39 3250 97.24 .1393 .8606 .1122 77.25 .0364 96.26 .2890 .8062 .0306 73.45 .1594 99.64 .1676 .9337 .1128 91.27 .0499 99.49 .1598 .9539 0798 94.19 0473 99.44 1459 .9241 .1992 91.50 1844 99.42 1965 .9184 0736 88.75 3515 86.24 3906 .4619 *.3270 47.29 5641 96.68 .9183 1573 .2317 92.17 .9555 91.22 .3075 .6357 4277 68.16 0249 94.51 .6510 .5728 .1286 76.85 .5385 92.68 .9102 1082 .2403 89.79 .9245 <	ts t 1 2 3 \$C* 1 2 97.61 6237 .3475 .6117 \$8.39 3250 .2509 97.24 .1393 .8606 .1122 .77.25 .0364 .8739 96.26 .2890 .8062 .0306 .73.45 .1394 .8437 99.64 .1676 .9337 .1128 .91.27 .0499 .9508 99.49 .1598 .9539 -0.798 .94.19 -0.473 .9653 99.44 .1459 .9241 .1992 .91.50 -1844 .8906 99.42 .1965 .9184 0736 88.75 5515 .8708 86.24 3906 .4619 -3270 .47.29 5641 .3830 96.68 .9183 1573 .231 .92.17 .9355 .0036 91.22 .3075 .6357 4277 68.16 0249 .6691 94.51 .6510

* Communality

92.70% of total variation is common.

Traits	Factor 1	Traits	Factor 2	Traits	Factor 3
14		5		. 17	
9		4		1	
12		6		19	口
13		2		22(neg.)	
15		7	\Box .		•
18		3			,
20		11			
21		10			
22	Ω	21			
8(neg.)	D				
11	Ω				

Figure 1. Comparison of size of loading for the variables included in the three main rotated factors.



racemes on the branches (4), and total number of branches (3); and for the negative, number of pods on the branches (5), average number of pods per productive node (10) and average long-internode length (19). Essentially this is a weight factor.

The second factor is highly associated with number-variables like number of nodes with pods on the branches (2), total number of branches (3), total number of racemes (6) and number of pods per plant (7), particularly with number of racemes on the branches (4) and number of pods on the branches (5), but also with number of pods per productive node (10) and average short-internode diameter (21) (Table 1 and Figure 1 on page 15). In general, the uppermost three or four short or basal internodes support most of the productive branches of the bean plant. Seed weight (9), pod weight (12), pod thickness (14) and pod length (15) have the lowest positive loadings whereas pod breadth (13) and long-internode length have the lowest negative loadings in this factor. Consequently, this factor indentifies itself with numbers, particularly number of branches, but not with weight nor with large pods.

The third and least important factor is characterized by loadings which reflect a numerous-thin-long internode relationship, that is, total number of nodes on the plant (1), number of long-internodes (17) and average long-internode length (19) have the highest positive loadings in this factor whereas the diameter-variables (20-22) along with average number of pods per productive node (10) and average pod length (15) have the highest negative loadings (Table 1, Figure 1 on page 15). Number of nodes with pod-bearing branches (2), number of racemes on the branches (4) and number of pods per plant (7) have the lowest positive loadings. Total



Table 2. Factor Analysis Results, East Lansing Set

			Unrotat	ed Factor	s			Rotated	Factors	
Trait	s 1 C*	1	2	3	4	1 C*	1	2	3	4
1	97.39	.5829	2427	.6629	2341	89.29	.1936	.1334	.9151	.014
2	97.66	.5788	.6758	.0551	.1095	80.67	.1251	.8778	.0448	.1360
3	95.95	.4990	.6385	.0225	.2026	76.46	.1107	.8397	0494	.211
4	99.77	.6386	.7000	.1141	1803	94.33	.0223	.9442	.2089	087
5	99.59	.6845	.7021	00+0	0481	96.39	.1485	.9658	.0877	037
6	99.73	.8129	.4811	.2218	0322	94.83	.2456	.8554	.3954	003
7	99.74	.8620	.4362	.0287	.0796	94.05	.4343	.8384	.2193	.028
8	91.10	.6354	073	3160	.5184	77.83	.8043	.2670	1484	.195
9	95.92	7878	.4763	.1621	2342	92.86	9289	0338	2501	.046
10	95.36	.5021	.6624	2395	1261	76.42	.0923	.8351	1351	200
11	93.41	.0378	.7870	.1314	.3516	76.17	2444	.6574	2308	.465
12	98.76	8175	.4147	.1828	.1549	89.77	7780	1191	3621	.383
13	94.59	8321	.2964	.2793	.0452	86.03	8120	2204	2191	.323
14	97.35	7655	.4142	:2667	1676	85.68	8967	0760	1629	.142
15	92.04	7060	.4067	0804	.2066	71.30	5790	0655	5411	.2840
16	84.90	.7913	0091	0582	1007	63.97	.5493	.4380	.3135	218
17	96.93	.5502	2615	.6748	1922	86.35	.1909	.0976	.9023	.057
18	85.08	6883	.2439	.3969	.1692	71.93	6801	1896	0947	.460
19	75.75	0983	.2380	.5455	.4970	61.10	1841	.1193	.1523	.7341
20	96.28	2282	.8564	1414	1634	83.21	5665	.5904	3983	062
21	97.40	2045	.8495	1254	2615	84.76	5906	.6023	3401	142
22	88.61	6009	.5502	3526	1600	81.37	6134	.1289	6297	155
E . D	nnte	8 7044	6 2745	1 0607	1 1502					

04 241 of total variation is common

Table 3. Factor Analysis Results, Gratiot Set

		14016	J. Tactor	74.41,515	McJures, orac			
		Un	Unrotated Factors			Rotated Factors		
Traits	\$ C*	1	2	3	1 C*	1	2	3
1	98.86	4887	.7250	.4063	92.95	3286	.3457	.8379
2	98.79	.4626	.7598	0133	79.15	.1391	. 8756	.0745
3	98.42	.6613	.5699	0485	76.45	.3209	.8041	1220
	99.76	.5690	.7705	.0237	91.80	.2390	.9260	.0579
Š	99.65	.6164	.7143	2501	95.28	.1289	.9461	2025
6	99.56	.2210	.9076	.1479	89.45	.0101	.8688	.3737
7	99.37	. 2052	.8667	2417	85.18	2204	.8944	.0576
8	91.77	1499	.5247	4556	50.53	5244	.4732	0804
ğ .	99.45	.8866	2781	.3175	96.42	.9373	.0969	2762
10	95.14	.6526	.0058	6041	79.08	.1394	.4032	7803
11	98.24	.8783	.3200	.0637	87.79	.6198	.6623	2347
12	99.11	.9135	1613	.2812	93.95	.9048	.2177	2710
13	95.45	.8034	2399	.3749	84.35	.8975	.0831	1766
14	97.68	. 8239	1879	.4388	90.67	.9367	.1266	1152
15	96.86	.8790	1746	.0629	80.71	.7535	.2291	4322
16	84.51	0957	.5511	.2996	40.26	0453	.3878	.5002
17 .	98.84	6014	.6539	.3801	93.38	4107	.2376	.8418
18	87.31	.5576	4545	.3038	60.98	.7264	2025	2027
19	93.22	6039	.3272	.5004	72.22	2534	0713	.8080
20	98.44	.9540	0339	.0366	91.28	.7576	.3888	4331
21	98.54	.9201	.2024	.0185	88.78	.6567	.5858	3366
22	95.34	.8110	2437	3129	81.50	.4989	.2041	.7241
	93.34	.0110						
E. Root	s	10.0829	5.8575	2.0810				
1 Trac		47.46	27.57	9.80	% Var.	31.80	29.94	20.18
Cum :		47.46	75.03	84.83	Cun. 1 Var.	31.80	61.74	81.92

. Communality



number of branches (3), number of pods on the branches (5), number of seeds per pod (8), average pod breadth and thickness (13, 14) and finally average short-internode length (18) have the lowest negative loadings in this factor. Therefore, this factor expresses a growth type where at the extreme an exagerate vegetative growth causes a very low production.

Factor Analysis of Individual Locations

The analyses by location show that 94.24% of the total variation is common at East Lansing, that is, the common factors accounted for that much variation (Table 2, page 17) compared to 96.56% at Gratiot (Table 3, page 17), indicating the presence of a slightly greater error at the first location. The results from these analyses were identical to those obtained with the complete set of data, with only one insignificant difference that the signs of the loadings in the first factor at East Lansing were inverted. That should not change the interpretation of the results, however. Another finding was the extraction of a fourth factor at East Lansing, confirming the fact already mentioned that there was more variance at that location. Moreover, the comparison between the four factors of East Lansing with the three of Gratiot gives the following correlations between the corresponding first three factors: -.93, .99 and .93 with no large values encountered for the fourth factor at East Lansing. But, the size of the correlations of that factor with the first and third of Gratiot (.33 and .37 respectively) may explain in part the relatively smaller correlations found for the first and third sets of factors (Table 3 a-b, page 19).

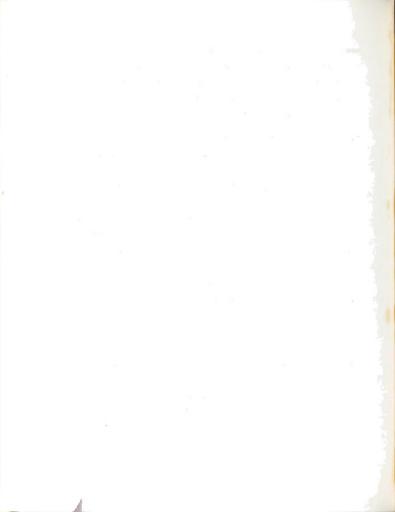
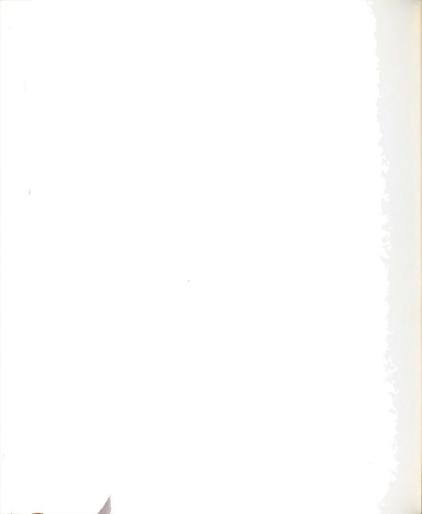


Table 3a. Factor Relationships Between
East Lansing and Gratiot Analyses
(Correlations)

East	G	ratiot	
Lansing	1	2	3
1	9315	.1307	.0329
2	.1147	.9894	0656
3	0915	.0498	.9264
4	•3329	.0390	.3692

Table 3b. Relationships Between Identical Pairs Of Variables From East Lansing and Gratiot Analyses

Variables	Correlations
1	•9476
2	.9858
3	.9268
4	•9559
5	.9571
6	.9739
7	• 9874
8	.8213
9	• 9451
10	.6710
11	.8962
12	.8686
13	.8931
14	.9328
15	. 8584
16	. 6379
17	• 9534
18	.9070
19	.3121
20	.9721
21	•9570
22	•9605



Factor Analysis of Determinate and Indeterminate Varieties as Sub-Sets

The results of the analyses by growth type show that there were four factors with roots greater than unity for the determinate lines and five factors for the indeterminate lines. There was greater error variance among the indeterminate lines than the determinate varieties as indicated by the amount of the total variation which is common in the two situations (92.01% compared to 95.63%). Here again, two major factors were recovered, but with some important changes in each case.

short internode length (18) but more on number of nodes with pods on the branches (2), on number of branches per plant (3) and also on number of racemes on the branches (4) (Table 4, page 21). The loadings for the three diameter traits (20, 21, 22) were markedly higher. The second factor was almost identical to that of the complete data set, with a little less importance given to the number of nodes with pods on branches (2) and number of branches per plant (3). These two factors account for more than 76% of the total variation before rotation and for about 64% after. The third factor was also recovered, but it can not be said that it is more important than the fourth factor of the bush-type structure, a factor which seems to contrast short internode length (18) with the number of short internodes (16) and also the number of seeds per pod (8).

For the indeterminate lines, the first factor was identical to the second factor of the complete set, whereas the second resembled the first factor of the complete set of data, but with two major differences. The average long internode length (19) has gained in importance and the diameter variables (20-22) were transposed from the second to the third



Table 4. Factor Analysis Results, Determinate Lines

	Unrotated Factors						Rotated	Factors	
Traits ! C*	1	2	3	4	\$ C4	11	2	3	4
1 98.46	6377	.5966	.3880	.1131	92.59	5310	.1235	.6672	4285
2 98.77	.5542	.7116	.3518	0412	93.90	.4388	.6992	.2354	4495
3 96.44	.6259	.6137	.3004	.0166	85.89	.5000	.6819	.1883	3293
4 99.56	.7489	.5847	0515	.1255	92.12	.4472	.8449	0431	0716
5 99.28	.6805	.6654	2089	.1578	97.44	.2943	.9357	1066	0273
6 99.15	.3063	.8690	1935	.1760	91.74	0718	.9393	.0687	1586
7 99.36	.1248	.8930	3367	.1464	94.80	2958	.9147	.0047 .	1540
8 91.90	2809	.5979	1887	5691	79.60	4588	.2632	2496	6737
9 98.73	.8939	3502	.1468	.1270	95.94	.9131	.0937	1557	.3045
10 93.63	.6369	.2014	6406	.1700	88.54	.1862	.6725	5078	.3751
11 96.62	.8139	.3825	0201	0810	81.58	.5810	.6442	2086	1408
12 99.14	.9223	2128	.1326	1493	93.58	.9065	.1462	3023	.0363
13 94.17	.8489	2759	.1671	0388	82.61	.8718	.0833	2066	.1283
14 98.14	.9010	2763	.0418	.0235	89.04	.8562	.1593	2774	.2344
15 96.20	.7822	2528	.2319	1839	76.33	.8453	.0147	2185	0288
16 89.42	1186	.5608	.3722	5113	72.84	0591	.1535	.1387	8259
17 97.77	6787	.4176	.3488	.3636	88.89	5470	.0407	.7540	1398
18 89.57	.4655	5677	.0333	.3109	63.67	.5486	1835	0590	.5465
19 77.44	4659	.1839	.3521	.5120	63.70	.3088	0172	.7278	.1076
20 97.53	.9538	.0600	.1120	.0702	93.09	.8397	.4459	1382	.0888
21 98.05	.9242	.1048	.1995	.0060	90.49	.8452	.4261	0925	0215
22 94.58	.8880	2276	.0469	0045	84.25	.8357	.1839	2751	.1862
E. Roots	10.6408	5.4162	1.5705	1.2988			•		
Trace	50.58	25.74	7.46	6.17	% Var.	38.03	26.14	11.02	10.84
Cum & Tr.	50.58	76.32	83.78	89.96	Cum & Var.	38.03	64.17	75.19	86.03
Total Trac									
* Comminal			5.63% of	total va	riation is	common.		•	

.

. Table 5. Factor Analysis Results, Indeterminate Lines

			Uni	otated F	actors				R	otated F	actors	
Traits	- 1 C4	1	2	3	4	5	\$ C*	. 1	2	3	4	5
1	82	.4444	.1540	.5118	0241	.0219	48	.2031	1796	.6225	0914	.1219
2.	98	.7608	.2408	4287	2300	.1171	89	.9087	0983	1076	0575	.1921
. 3	. 98	.7512	.2992	4011	2446	.1137	89 -	.9167	0478	0660	0696	.1888
4	100	.8735	.3463	2523	1000	1445	· 98	.9636	1398	.1604	.0199	0364
5	100	.9300	.2624	2064	0319	0463	98	.9335	2449	.1833	.0960	.0733
6	· 100	.8860	.3782	0540	0387	2081	98	.8976	1641	.3624	.0710	0827
. 7	100	.9517	.2114	.0145	.0683	0568	96	.8197	3455	.3759	.1381	.0832
8	93	.5894	1770	.0658	.1032	.7094	90	.2858	4120	.0975	.0872	.7927
9	96	6406	.6726	0011	.0185	1501	89 -	.2196	.8730	.0864	.0862	2457
10	95	.8483	.1533	0353	.0439	0450	75	.7393	3269	. 2774	.1121	.0763
. 11	93	.3503	.8359	.0196	0802	.1769	86	.6097	.5252	.3999	.0568	.2210
12	99	6053	.7391	0200	0998	.1780	95 -	.1608	.9587	.0559	0122	.0803
13		6969	.6297	0980	1719	.1258		.2269	.9336	0823	0851	.0099
14		6404	.6956	0884	1363	1929		.1404	.9213	.0043	0406	2957
15		3129	.6677	.3021	1291	.4627		.1110	.7287	.3585	1072	.4272
16	86	.6300	1562	. 3666	1639	0976	59	.3192	5030	.4281	2311	.0213
17	78	.1601	.1172	1738	.7736	0130	67	.0971	0660	.0370	.8082	.0048
· 18		5157	.6355	1651	.0150	1942	73 -	.0644	.7975	0303	.1189	2825
19	73	2993	.4626	3857	.4402	.1281	66	.0178	.5490	2049	.5618	.0564
20	90	.4020	.6649	.3096	.1979	1923	78	.4372	.2424	.6750	.2393	1112
21	94	.5611	.6091	.3483	.2264	0196	86	.5033	.1194	.7194	.2575	.0859
. 22	84	.1024	.2946	.6269	.0193	0761	50 -	.0563	.0839	.6935	0702	0168
F.1	Roots	8.8850	5.1820	1.7811	1.1354	1.0615						
		13.90	25.60	8.80	5.61	5.24 1	Var	30 18	27.41	12.77	5.77	5.60
		. 43.90	69.50	78.30	83.91		Cum.	-0.40	~·· • • •	10.,,	J.,,	3.00
		ace = 2						30.48	57.89	70.66	76.43	82.03
		ality		92.0	1% of to	tal vari	ation	is com				



factor (Table 5, page 21). Consequently, the third factor becomes an association between the number of nodes (1) and the diameter variables (20-22) instead of the number of long internodes (17) as in the complete-set results. It is a significant change per se, because this third factor extracts here about the same amount of variation as it does in the complete-set analysis. The fourth and fifth factors each accounted for less than 6% of the total variance. Number and length of long internodes (17, 19) had high loadings in the fourth factor, and number of seeds per pod (8) showed a high loading in the fifth factor. These variables had also low communalities in this analysis (Table 5, page 21). The first two factors extracted a little more than 69% of the total variation prior to rotation and about 58% after rotation.

Comparison Between the Factor-vectors Extracted in Analyses of Determinate and Indeterminate Varieties

The factor-variables from the determinate and the indeterminate lines were compared and the results show, in addition to the transposition between the first and second factors of the two sets, that factor #3 of the determinate group is similar to factor #4 of the indeterminate group and also that factor #4 of the former is the opposite of factor #5 in the indeterminate group (Table 5a, page 23). Consequently, factor #3 of the indeterminate sub-set is unique to that group and can be seen as another expression of the third factor of the complete-set analysis. The following variables in decreasing order of importance account for most of the differences between the determinate-group and indeterminate-group results: number of nodes on the plant (1), average number of pods per productive node, (10) average long internode diameter



Table 5a. Factor Relationships Between
Determinate and Indeterminate
Lines Analyses (Correlations)

		Bus	h	
Vine	· 1	2	3	4
1 2 3 4 5	.0498 .8647 .4740 .0087	.9650 1197 .1616 .0945 1399	0765 0547 .0223 .9744 .2029	.0025 .4007 4834 .1908 7546

Table 5b. Relationships Between Identical Pairs of Variables From Analyses of Determinate and Indeterminate Lines

Correlations		
.1728		
.6613		
.6944		
.8817		
.8985		
.9783		
.9365		
.7253		
.8808		
•4399		
.9412		
.8049		
.8015		
.8442		
.9105		
.6602		
.7778		
.9875		
•4639		
.7738		
.7376		
.4600		



Table 5c. Multiple Discriminant Analysis between Determinate and Indeterminate Lines

Variables	Discriminant Axis Loadin	gs
1	9418	
2 3	.1810	
	.0003	
4	.3076	
5	.1535	
6	.4906	
7	.2750	
8	.1289	
9	2645	
10	1866	
11	0694	
12	2511	
13	2139	
14	1413	
15	3661	
16	.3882	
17	.8727	
18	2497	
19	.3359	
20	 3193	
21	1638	
22	5614	
% of Total Variation :	100.00	
Chi-Square Value :	317.593	
Degrees of freedom :	22	
Significance Level :	.0000	



(22) and average long internode length (19) (Table 5b, page 23).

These results were also confirmed by the use of multiple discriminant analysis. Total number of nodes on the plant (1) was the main differentiating variable between determinate and indeterminate lines, followed by the number of long internodes (17) and also their diameter (22) (Table 5c, page 24). The discriminant function had a highly significant chi-square.

Factor Analysis for the Three Seed-weight Groups

Attempts were also made to test the invariance of the factors when, instead of classifying by growth type, the material was divided into seed-weight sub-sets. The average seed weight before planting was used for this purpose. Three groups were established; the light-seeded, the medium-weight and the heavy-seeded lines. They corresponded approximately to the navy, the great northern and the kidney beans with an average weight of 15-16 gms per hundred dry seeds, 29-30 gms and 40-45 gms, respectively.

Factor analysis for the light-seeded lines reveals that 93.61% of the total variation was common, and that 89.67% was extracted by 5 roots (Table 6, page 26). The first factor accounted for 46.20% of the trace before rotation and 41.6% after rotation. It was by far the most important factor, since the next two factors together extracted only some 31% of the trace. That factor resembles a combination of the first two factors extracted or identified from the complete set of data. The short internodes and hypocotyl diameters of the first factor were combined with the number-variables of the second factor of the complete set to form the first factor of the light-seeded lines. Three remarks should



Table 6. Factor Analysis Results, Light-Seeded Lines

	Unrotated Factors							Rotated Factors					
Tra	its 1	C*	1	. 2	3	4	5	\$ C*	. 1	2	3	4	5
•	1	98	.7193	4807	.1565	,2863	2584	92	.7425	2362	.2849	.4619	1418
	2	98	.8437	0300	4062	.1763	.0639	91	.8473	.1617	3523	.2091	0326
• • •	. 3	98	.8053	.0584	4352	.0967	.1139	86	.8023	.2085	4067	.1005	0308
	.4 1	00	.9614	1634	0179	0063	.0480	95	.9715	0128	.0121	.0258	0943
	5	99	.9585	1015	0476	0923	.1870	97.	.9785	.0151	0542	1172	0181
	6	99	.9581	0718	.1304	0417	.0662	95	.9527	.0577	.1426	0721	0987
		99	.9323	.0285	.0497	1636	.2211	95	.9340	.1043	.0157	2526	0287
		95	.3892	.5188	6513	.0593	0718	85	.2970	.5501	6373	.0494	2335
		91	.1406	2468	.8573	.1942	1258	87	.1451	1359	.9037	-1082	.0365
		93	.5943	.0450	.0848	5285	.3671	78	.6176	0624	0381	6242	.0086
		95	.7473	.5557	.1233	0241	.1600	91	.6590	.6107	.0790	2977	0804
		97	.2318	.8367	.2408	.3157	0495	91	.0819	.9159	.2439	0438	0826
			0125	.6510	1494	.5035	.1737	73	0828	.7888	1344	.1602	.2391
• •		93	.2115	.2923	.7949	.2847	.0090	84	.1485	.4054	.8052	0358	.0841
		90	.1643	.7994	.0053	0817	1870	71	.0008	.7179	0271	2483	3604
		93	.6887	.1100	1000	0620	4935	74	.5814	.1608	0240	.1401	5997
		98	.5692	6.258	.1652	.3592	0296	87	.6546	3560	.2770	.4706	.1394
			5787	.1110	.5428	1551	.1087	68	5789	0365	.4662	3217	.1434
. •			2093	1830	.1249	.3788	.6237	63	0788	0354	.0969	.0798	.7760
		97	.8828	.0462	.3324	1244	1075	92	.8292	.1184	.3420	1472	2806
` .		97	.9133	1399	.2238	0107	0656	91	.8892	0032	.2605	.0113	1778
	22	89	.0865	.2992	.6120	3491	.0261	59	.0228	.1549	.5220	5249	1467
•	E Po		9.5144	3.3311	3.1773	1.3594	1.0856						*
•	\$ Tr			16.17	15.43	6.60			.41.63	14.42	14.15	7.19	6.54
			r 46.20	62.37	77.80	84.40	89.67	Cum	.41.05				
			race = 3		,,,,,,	57.70		Var.	41.63	56.05	70.20	77.39	83.93
					93.6	51% of to				mmon.		•	•
	e Co	Time U	nality										

Table 7. Factor Analysis Results, Medium-Weight Lines

		Unrotate	ed Factor	s ·			Rotated	Factors		
Taits & C *	1	2 .	3	44	\$ · C*	1	. 2	3	4	
1 99.10	7367	.5662	.2964	.1021	96.16	9312	.0133	2691	.1478	
2 98.98	.6149	.6719	.1374	1794	88.06	.1056	.9248	1140	.0356	
3 98.57	.7203	.5503	.1182	1754	86.64	.2452	.8966	0367	.0316	
4 99.94	.6111	.7291	.1424	2424	98.40	.0719	.9799	1354	0158	
5 99.91	.7053	.6779	.1096	1181	98.30	.1900	.9595	1266	.1007	
6 99.88	.3600	.8531	.2814	1702	96.54	2334	.9379	1631	.0689	
7 99.82	.5239	.7858	.2142	.0175	93.81	0339	.9219	1644	.2452	
8 93.62	.4252	.4260	.1197	.6924	85.60	.1379	.4114	1419	.8047	
9 97.80	.6121	7136	.0678	0413	89.02	.7185	0995	.6016	0460	
10 98.25	.7634	.4643	1518	0390	82.30	.4596	.7484	2063	.095	
11 97.89	.8982	.1093	.3252	.0807	93.09	.4758	.6913	.3831	.282	
12 98.95	.5973	7027	.2312	.1656	93.14	.6360	1056	.6932	.187	
13 95.06	0395	7566	.5534	.0053	88.03	.0024	3905	.8530	.015	
14 97.59	.2492	7481	.3299	4106	89.92	.2985	1706	.7916	3929	
15 98.71	.8185	4243	0865	.2908		.8584	.0964	.3204	.305	
16 94.56	2755	.7301	2037	.2927	73.61	3835	.2093	6736	.302	
17 99.44	7423	.4771	.4253	.0048	95.96	9712	.0104	1066	.070	
18 77.35	.2738	5415	.5018	.0189	62.03	.1778	0707	.7589	.0884	
19 87.04	3858	.0529	.7280	.2516	74.49	6615	0453	.4196	.359	
20 97.98	.9090	2867	0392	.0597	9182	.8312	.3121	.3251	.155	
21 98.79	.9424	1163	0442	.1132	91.63	.7906	.4436	.2287	.205	
22 98.21	.8468	25-9	2892	.0438	86.92	.8951	.2330	.1054	.051	
E-Roots	9.0322	7.3545	1.9931	1.1172				•		
1 Trace	42.46	34.57	9.37		: Var.	31.43	32.22	18.50	6.43	
	42.46	77.03					63.65	82.15	88.6	
Cum. & Tr.			86.40	31.04	Cum & Var.	31.43	03.03	04.13	00.04	
Total Trace	= 21.27		5 709 a.f		min e ion in					
		91	0./US OF	total va	riation is	common.			•	

* Communality



be made here. Total number of nodes on the plant (1) was included, so was the number of long internodes (17). The pod dimensions (12-15) along with the average long internode diameter (22) vanished.

Finally, the number of short and long internodes (16, 17) had positive loadings instead of negative loadings as in the complete-set results and the opposite occurred for the average short internode length (18). So, this is a general factor pointing toward a plant type with sturdy, short internodes producing many branches full of productive units. There can be many nodes on such a plant. The second factor in the light-seeded sub-set was a pod factor characterized by long, wide and filled pods (12, 13, 15) leading to a relatively heavy plant (11). The third factor expressed a contrast between seed size and weight (14, 9) and seed number per pod (8). Such a type would probably have sturdy long-internodes (22) and also long short-internodes (18).

In the analysis of the medium-weight set, there was a relatively smaller error variance, 96.70% of the total variance was common, and 91.64% was extracted by 4 roots. The first two vectors were about equally important as indicated by the amounts of variance extracted by them, 42.46% and 34.57% before rotation and 31.4% and 32.2% after rotation, respectively (Table 7, page 26). The second factor was identical to the second factor of the complete set of data. The first, however, was apparently a completely new factor. It is bipolar and has large positive loadings on the diameter-variables (20-22), the length and the weight of the pod (15, 12) and seed weight (9), and negative loadings particularly on the total number of nodes (1) and on the number and the length of the long internodes (17, 19). Number of pods per productive



node (10) and average plant fresh weight (11) have positive loadings but intermediate in size. The third factor contrasts the number of short internodes (16) with seed weight (9), the pod dimensions (13-15) and the average short internode length (18). The number of short internodes (16) had negative loading. This factor extracted 9.37% of variance before rotation and 18.5% after rotation. It was the most important third factor encountered in this study (Table 7, page 26).

The analysis of the heavy-seeded sub-set shows that 95.64% of the total variation was common and 89.0% was extracted by 5 roots. Only the first three deserve consideration, however (Table 8, page 29). The first factor was essentially the same as the second factor of the complete set analysis and the second was comparable to the third of those factors, but with some important differences. The negative loadings of the diameter-variables (20-22) and number of pods per productive node (10) were significantly higher in this factor. Finally, the third factor is identifiable with a contrast of plant weight (11), pod weight (12), pod breadth (13) and pod thickness (14). There were no high positive loadings on this factor. The three factors-variables of this set accounted for 35.59%, 26.89% and 11.15% of the total variation before rotation and 26.36%, 24.52% and 14.17% after rotation, respectively.

<u>Comparisons</u> <u>Between</u> the Factor-Vectors <u>Extracted</u> from Analyses of the <u>Seed-Weight</u> <u>Groups</u>

The three sets of factors from the three seed-weight groups were compared two by two. Through these comparisons, it can be seen that the second factor of the complete-set factors was found in the results of all three seed-weight groups, although with some modification in the

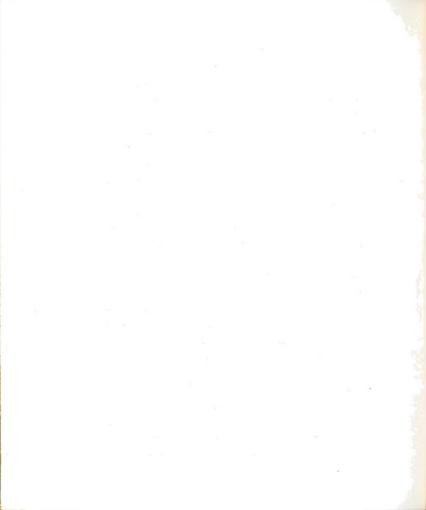


Table 8. Factor Analysis Results, Heavy-Seeded Lines

Unrotated Factors						Rotated Factors							
Tra	its	\$ C*	1	2	3	4	5	% C∗	1	2	3	4	5
	1	99	1809	.9505	1302	.0722	.1396	98	.2478	.9286	.0711	.1438	.1685
	2	98	.5940	.6112	.2779	.0329	.1036	82	.6792	.1856	0855	.2401	.5047
	3	93	.6702	.4152	.4546	1989	.0390	87	.6616	0839	.0101	.0649	.6482
	4	100	.7870	.4801	1473	0697	3118	97	.9638	.0671	1614	1037	.0583
	5	100	.9126	.2500	0802	.0142	2782	98	.9374	1889	2491	0176	.0507
	6	100	.6619	.6499	0885	.0264	2457	93	.9193	.2510	0928	.0326	.1080
	7	100	.8265	.3880	0074	.1363	1823	89	.8945	0458	2176	.1512	.1143
	8	90	.1331	0343	.1308	.8614	11178	79	.0566	0930	1463	.8471	2022
	9	94	.567 6	4943	0668	4653	.0509	7,9	.1632	6100	3758	4678	.1764
]	10	98	.7239	3536	2165	.0559	3527	82	.5537	5752	2954	1183	2914
1	11	96	.8308	.1840	.0637	.2443	3429	91	.5649	1381	5634	.3612	. 3453
	12	95	.6464	2511	1481	.2155	.5577	86	.1308	3067	8082	.2478	.1857
1	13	89	.5730	3017	4987	0379	.3172	77	.1638	2:58	8043	1674	0895
	14	96	.6105	0649	6634	.0320	.2511	88	.3400	0202	8397	1521	1916
	15	94	0058	4777	.7690	0037	.1534	84	3209	6301	.2635	.2607	.4536
	16	94	0102	.4424	.4712	.5224	0966	70	.2449	.1924	.3638	.6718	.1392
1	17	99	238 5	.8805	0719	1066	.2748	92	.1019	.8973	.0478	.0232	.3252
	18	89	.2699	.2419	.3271	6046	.1522	63	.2164	0039	.0573	3553	.6714
	19	93	0498	.7532	.1722	2780	.3674	81	.1298	.6463	.0337	0290	.6126
	20	97	.776 7	3495	.3779	1769	0765	91	.4427	7512	1377	0479	.3517
	?1	97	.7345	4315	.2986	.0765	.0398	82	.3 309		2753	.1641	.2248
	22	94	.3627	8042	.1951	.0490	.1470	84	1706	8479	2909	.0705	.0525
I	. Ro	ots	7.4877	5.6581	2.3454	1.9070	1.3286						
	Tr		35.59	26.89	11.15	9.06	6.31	& Var	.26.36	24.52	14.17	9.06	11.02
			35.59	62.48	73.63	82.69	89.01		26.36	50.88	55.05	64.11	75.12
								% Var		22.00		- · •	
Total Trace = 21.0401 Communality					95.6	4% of to	tak var		-	mon.		•	

Table 6a. Relationships between Factors of the Light and Heavy-seeded lines (Correlations.)

ight-see	d ed		Heavy-seeded		
•	1	2	3	4	5
1	.9570	.0551	1064	.1785	.1946
2	1801	2833	7882	.5115	.0660
3	0311	2918	2381	6274	.6808
4	2109	7353	.0184 ·	.2869	.5764
5	.0784	.5394	5571	4801	-,4025

Table 6b. Relationships between identical pairs of variables from Analyses of Light and Heavy Sets.

Traits :	1	Correlations	: .5	767
	2			3381
	3			
	Ž.			457
	7			201
	3			515
		•		067
•	<u>'</u>			873
	8			895
•	9		.6.	562 ·
	10	,	.9	114
	11			166
	12 .			
	13		7	558
	14			250
	15			684
	16			459
	17			276
	18			040
	. 19			181
	20			379
	21	•		
•	22			640
	44		2	233



light-seeded results (Table 6 a-b, page 29). This confirms the finding that the first factor of the complete-set analysis is indeed a weight factor, since in none of the seed-weight sub-set analyses was it possible to recover it. These comparisons also indicate that the fourth factors of the medium-weight and heavy-seeded groups are almost the same. Furthermore, combinations of factors in one set may probably have the same effect of another factor or combination of factors in another set, an exception being made of the factors similar to the second factor of the complete set results (Table 7 a-b, 8 a-b, page 31).

These results were verified through multiple discriminant analysis. Of the two discriminant functions necessary to separate the three groups, both with very significant chi-squares, the first was the most important. It accounted for 88.60% of the variance (Table 8c, page 32). Seed weight (9), and the pod-dimension variables had the highest positive correlations with that function. This function is almost identical to the first factor-variable found through factor analysis of the complete set of data. The second function, on the other hand, resembles more the third factor from that structure. It is also similar, in pattern, but not in size of the loadings, to the discriminant function between vine and bush types, that is, it has high correlations with number of nodes per plant (1) and number of long internodes (17). These results show also that the three seed-weight groups can include branchy types of plants.



eight 4 3 4 08321731	'	cen Identica From Analyss d Medium-	Correlations	9336	.8101	1866.	9665	.7345	.9011	.9103	.7828	.5858	. 90.12	.9521	8728
Mcdium-Weight 2 3 .9810083	0632 1002 1538	Relationships Between Identical Pairs Of Variables From Analyses Of Heavy-Seeded And Hedium- Weight Lines	•												
10008	.1270				•										
Heavy- Seeded	1640	Table 8b.	Variables		۲ ۲	31	n 🐠 i	~ &	6 5	2 = 1	12 13	14	. 1 E	17 18	19

Correlations

Variables

.2834 .8427 .7824 .9850 .9880

Relationships Between Identical Pairs of Variables From Analyses of Light-Seeded and Medium-Weight Lines

Table 7b.

.0755 .7388 .6218 .2456

-.0607 .5254 .6005 .0543

.0710

Relationships Between Factors Of Light-Seeded and Medium-Weight Lines (Correlations)

Table 7a.

Medium-weight

Lightseeded

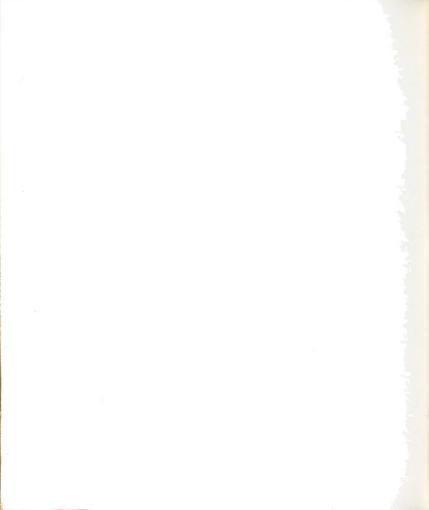


Table 8c. Multiple Discriminant Analysis between Light-seeded, Medium-weight and Heavy-seeded Lines

Variables		Discrimir 1	ant Axes Loadings 2
1		2864	.5705
2		.0493	0356
3		.1351	1261
4		.0439	.0145
5 6		0634	0462
6		2134	.1899
7		4005	.1214
8		5265	.2964
9		.9115	1776
10		0127	1501
11		.4259 ·	.0305
12		.8874	0830
13		.9159	1644
14		.8987	1240
15		.7612	.0784
16		2884	.1546
17		3580	.5227
18		.6435	3102
19		-11154	.1257
20		.5787	1801
21		.5693	0167
22		\$5543	1267
6 Of Total Variance	:	88.60	11 .40
Chi-Square value	:	292.655	105.486
Degrees of Freedom	:	23	21
Significance Level	:	.0000	•0000



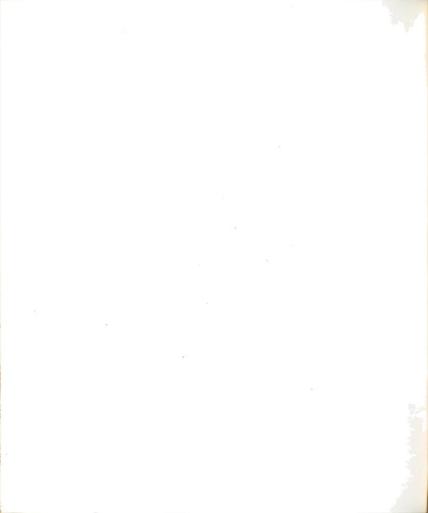
DISCUSSION

Interpretation of the results

Factor analysis of the complete-set of data reveals that there are two major factors or patterns of production in beans and a much less important factor more related to the morphology of the plant than to its productivity. Thus, the relation of this third factor to production is indirect. The three factors have been tentatively called pod-weight, pod-number and growth, respectively. All three factors remained invariant from one location to another despite the different types of planting which were used to accentuate the differences between the two environments. It is to be noted, however, that that the means of many of the variables were found to be significantly different between locations.

Factor analyses with the growth types show that these two major factors exist among both determinate and indeterminate lines and furthermore, that whereas the first factor is the most important for the bush type, the second is probably the main factor for the vine type. The third factor was also recovered in both cases, but with a different configuration among the indeterminate lines. Consequently, it can be concluded that these patterns are consistent across locations and growth types.

Multiple discriminant analysis, both between growth types and between seed-weight groups, establishes beyond any doubt that the third factor of the complete-set results is indeed a growth factor. It was



responsible for about 10% of the total variation before rotation and 15% after rotation in the factor analysis of the complete-set of data. This indicate that growth habit has a relatively low impact on bean productivity. The factor of growth can be expressed in terms of the total number of nodes and the number of long internodes on the plant. Probably, the greater the number of internodes of the plant (presumably indeterminate), the thinner the stem and the pods and also the lighter in weight will be the whole plant. In an extreme case, the plant will certainly produce too many racemes, many undeveloped pods giving way to an unbalanced set of yield components, and consequently a very low yield.

The results of the factor analyses of seed-weight groups data are in accord with those of the complete-set. The first factor in the latter was in fact a weight factor, since it was not recovered in any of the seed-weight groups analyses. But, the second factor, or number-factor was identified in all of them. This indicates that the grouping of bean varieties by seed weight is very appropriate and its implication on productive potential is great, given the importance of the weight-factor in this study.

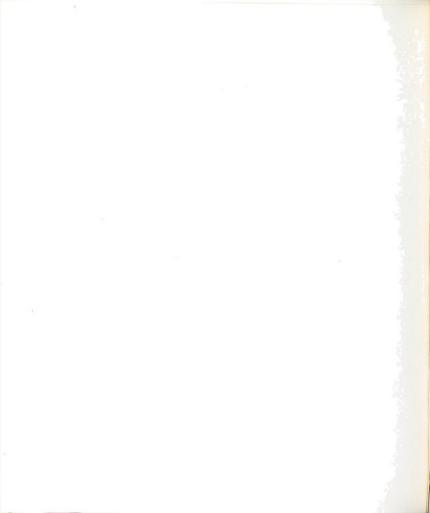
As a multiple discriminant function, the weight factor accounted for about 89% of the total variance. In the results from factor analysis of the complete-set of data, it was as important as the number-factor. This number-factor was recovered in all seed-weight group analyses and as in the indeterminate sub-set, it was the most important factor in the light-seeded sub-set results.

Implications of the results on the Development of an Improved Plant-type in beans

The results indicate that in order to reach a high yield through high seed weight, a plant has to have a sturdy stem and relatively few but long basal internodes. Such a plant can not therefore also have many nodes and certainly will have no or very few long internodes. It will be more feasible to increase the number of pods on that plant than to augment its number of seeds per pod. Its production will be mainly or uniquely located on the stem. The height of the plant is due mostly to the length of its basal internodes.

By contrast to this pod-weight type, in order to attain a high yield per plant through a high pod number, a bean plant needs many basal internodes which should be moderately sturdy. The number of nodes on the plant can be higher than in the preceding type. Its pod-bearing potential is higher on the branches than on the stem. This plant-type can have considerably more seeds per pod, but the seed-weight should be kept at a minimum. Interestingly, this pod-number type can weigh more than the pod-weight type and is also more efficient since its number of pods per productive node is higher. The height of the plant is due primarily to the number of basal internodes.

The factor related to growth shows that a bean plant with numerous, long and thin internodes will be an extremely poor yielder. This also will be a strong competitor. Consequently, in improving yield, either on a per-plant basis or on an area basis, such a plant-type should be avoided. However, this result can also be seen as an invitation for inter-breeding between the pod-weight and the pod-number



types, so long as it is possible to maintain a favorable balance between the number of nodes and the diameter of the stem. Whenever this can be done and at the same time the number of upper-internodes can be kept at a minimum, the yield of the plant, whatever its original plant-type will be improved considerably.

It is interesting to note that although most of the unadapted, south-american varieties were not included in the factor analyses, their yields and morphological characteristics strongly confirm the result that too many nodes or internodes on the plant, particularly the upper ones, will provoke a reduction in its yield,

It seems therefore feasible to attempt to develop an improved bean plant-type by increasing the mod-bearing capacity of the stem of a heavy-seeded line. This can be done by augmenting the number of its short or basal internodes and also maintaining the sturdiness of its stem. If the seed size can be kept at a level commercially acceptable, without increasing the number of branches, this superior planttype will fit well to the objective of improving yield on an area basis, because inter-plant competition will be thus minimum. However, the podbearing capacity of the plant as a whole can still be enhanced by augmenting the number of its branches. This can be realized through backcrossing with a pod-number type parent. This may not appeal to most plant breeders in view of the fact that the inter-plant competition might be consequently increased. But, it still may be wise to do it, at least as an intermediate step, since this is likely to produce a reduction in seed size, something which in itself may be economically beneficial. Moreover, the pod-number type parent may also be used as



as the carrier of other useful traits. But, if the objective is to improve yield on a per-plant basis, the number of branches on the plant should always be a matter of concern, since the higher their number, the higher the yield. In fact, everything else being comparable, any bean plant will yield more with than without branches.

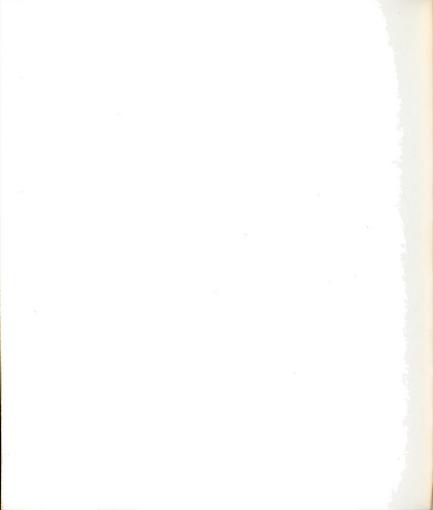
So, keeping in mind the limit imposed by the third factor, the growth factor, several improved bean plant-types can be obtained through repeated inter-breeding between the two major plant-types, namely the pod-weight and the pod-number types, Shifting emphasis, through increasing the number of backcrosses toward one type or the other, permits the plant breeder to enhance the differences between those improved plant-types, that is, to separate genetically the ones approaching his ideal of a good variety on a plant-yield basis from those getting close to his "ideotype" on an area-yield basis.



SUMMARY AND CONCLUSION

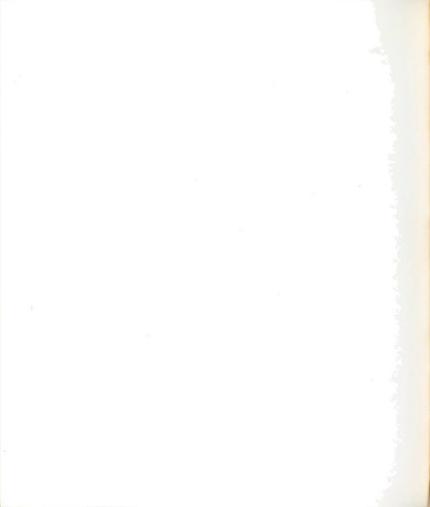
There results of this study indicate that there three uncorrelated factors or patterns of production in beans and that only the pod of the plant can be used to distinguish between the first two. The first most important factor is pod-weight, the weight of the pod as expressed through the size of both the seed and the pod. The second most important factor is pod-number and is related more with the number of racemes on the plant than with the number of pods per productive node. It is interesting to note that the number of branches on the plant has only a very low weight in the obtaining of this factor. Surprisingly, number of seeds per pod does not have a high positive loading on any of these two major factors. On the contrary, that variable has a negative and moderately high loading only in the first factor. The third and least important factor is a growth factor with emphasis on the general size of the plant. Incidentally, this factor will not necessarily discriminate between determinate and indeterminate lines. Its resemblance with the discriminant function calculated for that purpose indicates rather that the indeterminate lines are more likely to have many nodes and long internodes as compared to the determinate lines.

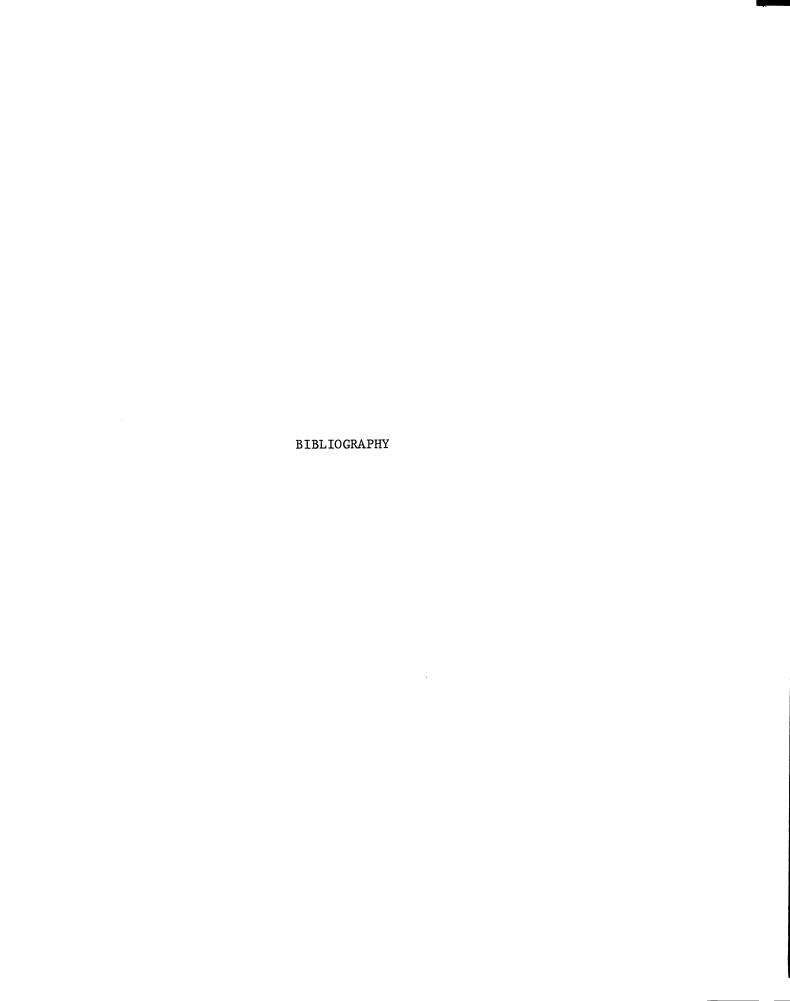
It was found that it is possible to make progress in yield improvement by breeding within any of the two major types, that is, increasing the weight of the seed of a pod-weight type variety or the number of pods of a pod-number type variety will often result in better yields. In the first case, the short or basal internodes



should be made longer and sturdier whereas in the second they should become more numerous while remaining sturdy. However, greater advance is obtainable through interbreeding between the two types. For example, the number of short internodes can be increased in a pod-weight type line so that it can produce more pods. In the same manner, the weight of the pod and the seed can be improved in a pod-number type variety by making its short internodes longer and sturdier. However, two variables should be kept under control in either case, the number of long internodes and the number of branches. The first will prevent the plant from approximating its maximum yield potential, and the second will make the plant a strong competitor, thus limiting its use to production on a per-plant basis.

Obviously, since beans are categorized for commercial usage largely on the basis of seed size, the plant breeder therefore must place greater emphasis upon obtaining a plant-type which, on an area basis, maximizes the pod number.







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