# OPTIMUM PLOT SIZE FOR FIELD EXPERIMENTS ON TURMERIC (CURCUMA DoNGA. L.) 



THESIS
Submitted in partial fulfilment of the requirements for the degree of Taster of Science (TAgriculfural \$tatisfics) Faculty of Agriculture Kerala Agriculture University

Department of Statistics

Mannuthy. Trịchur-


I hereby declare that this thesis entitled Optimum Plot Size for Field Experiments on Turmeric (Curcuma: Longa. L.) is a bonafide record of research work done by me during the course of research and that the thesis has not previously formed the basis for the award to me of any degree, diploma; assciateship, fellowship, or other similar title, of any other University or Society.

Mannuthy,
16-8-1984.


## CERTIFICATE

Certified that this thesis entitled Optimum Plot Size．for Field Experiments on Turmeric（Curcuma， Longe．I．）is a record of research work done indepen－ dently by Sri．Gopakumaran Nair．B．，under my guidance and supervision and that it has not previously formed the basis for the award of any degree，fellowship or associateship to him．

## ACKNOWLEDGEMENT

I hereby acknowledge with deep sense of gratitude the help rendered by the supervising teacher Sri.P.v. Prabhakaran, Professor and Head of Department of Agricultural Statistics, College of Agriculture, Vellayani, who guided the work, besides being constant source of encouragement to me.

I take this opportunity to express my sincere thanks to the members of the advisory committee namely, Dr.c. Sreedharan, Professor of Agronomy, Sri.s.Balakrishnan, Professor of Horticulture (Farm) and Sri.V.K.Gopinathan Unnithan, Associate Professor of Agricultural Statistics, College of Horticulture, Vellanikkara to give me the time to time advice.

I like to express my gratitude to the staff members of the Department of Plantation Crops especially, Sri.Nybee, Assistant Professor, College of Horticulture for the help rendered during the field trials.

I am grateful to the Kerala Agriculture University for the award of funior fellowship during the course of the study.

I have great pleasure to acknowledge the unstained help and assistance of my colleagues and staff members of Department of Statistics, College of Veterinary and Animal Sciences, friends and all others whose timely help and encouragement enabled me to fulfill this venture.

Mannuthy,
16-8-1984.


1. INTRODUCTION ..... 1
2. REVIEW OF LITERATURE ..... 9
3. MATERIALS AND METHODS ..... 24
4. RESULTS ..... 39
5. DISCUSSION ..... 66
6. SUMMARY ..... 76
7. . REFERENCE
8. APPENDICES
9. ABSTRACT
SI.
No.
Title of the Table ..... Page
10. Observed coefficient of variation for plots ofdifferent sizes when plots are not grouped49
11. Observed coefficients of variation for plots of different sizes when plots are grouped into different sizes and without grouping ..... 50
12. Estimated coefficients of variation for plots of different sizes in blocks of different sizes and without blocking as obtained from,
a) Smith's equation, $y=a x^{-b}$ ..... 51
b) Mode1, $Y=a+b / \sqrt{x}+c / x$ ..... 51
13. Optimum plot sizes as estimated from Smith's equation considering different cost ratios ..... 52
5 \& 6.Estimates of parameters, coefficient of deter- mination and F-ratios for various models fitted to the data with blocks of different sizes and without blocking,
5a) Smith's model, $Y=a^{-} \mathrm{b}^{\prime}$ ..... 53
5b) Model, $y=a+b / \sqrt{x}+c / x$ ..... 53
6a) Model, $Y^{-1}=a+b$ log; $x$ ..... 54
6b) ModeI, $\mathrm{y}^{-1}=a+b \sqrt{x}+c x$ ..... 54
14. Percentage relative efficiencies of plots of different sizes in blocks of different sizes and without blocking ..... 55
15. Optimum plot sizes estimated from the Smith's equation by modified maximun curvature method ..... 56
16. Effect of size and shape of plots on coefficients of variation for blocks of different sizes and without blocking ..... 57
S1.
No.
Title of the Table
17. Percentage relative efficiencies of blocks of different sizes compared with no blocking for plots of different sizes ..... 58
18. Percentage relative efficiencies of blocks of different shapes compared with no blocking for plots of different sizes ..... 59
19. Relative efficiencies of blocks of different sizes without considering the sizes of the plot using Smith's variance law ..... 60
20. Minimum number of replications and total experi- mental area required for estimating treatment m means with 5\% standard error ..... 61
21. a) Percentage relative efficiencies of Latin Square Design over Randomised Block Design ..... 62
b) Percentage relative efficiencies of Latin Square Design over Completely Randomised Design ..... 62
22. Percentage relative efficiencies of Randomised Block Design over Completely Randomised Design ..... 63

*     *         *             *                 *                     * 

LIST OF ILLUSTRATIONS
S1.
No. Title of the Figure Page
No.

1. Fertility contour map ..... 64
2. Effect of plot size on varlability ..... 65

INTRODUCTION

## INTRODUCTION

Statistical methodology plays an important role in evolving appropriate agrotechniques for the enhancement of crop production. The formulation of proper methodology for collection of data, their analyses: and interpretations help In this regard. As is well known, field experimentation is the most powerful tool of agricultural research and it can be successfully conducted if and only if the experimenter has got some idea regarding the variability of the experimental material. There are two principal sources of variation in field experiments. They are (i) variation due to soil heterogeneity and (11) variation due to inherent variability (genetic variability) within the crop species.

These two types of variabilities are inherent in any experimental material and because of their inheritance, it has become difficult to compare the differences between treatments. Even if treatments are different in their effect, no one is sure as to whether the differences are due to the treatment effects or due to inherent variation in soil heterogeneity. Thus the outcome of any biological experiment becomes a stochastic variable and statistical principles based On the laws of probability are to be applied in the study of such phenomenon. Plot-to-plot variation due to uncontrolled factors such as soil fertility is generally called experimental
error and if left uncontrolled, it can offmset experimental findings.

The basic principles of the theory of experimental designs involving the well known concepts of replication, randomisation and local control were originated by Fisher (1926) during the course of his experimental work at the Rothamsted Experimental Station between the years 1921-25.

With the introduction of these principles, field experimentation was based on a scientific footing and methods 'of logical construction of the experiment were knewn to the experimenter enabling him to draw objective ${ }^{(1)}$ and reliable conclusions with pre-assigned degree of precision. of these three principles, replication and local control were meant for reducing variation and improving precision of the estimates. Randomisation along with replication provided a valid estimate of error variance.

All these procedures are collectively called "The Direct Methoas" of controlling error and are distinctively different from the statistical control of error through analysis of covariance. . The direct methods of controlling error include in addition to replication and local control, such devices as selection of uniform site for experimentation, provision for border rows to eliminate border effect, maintaining uniformity in the physical conduct of the experiment,
replanting of dead hills or missing plants, eliminating offtypes, controlling the incidence of pest and dicease, proper orientation of plots and biocks and adoption of an optimum size and shape of plots and blocks for the conduct of the experiment. of these, the simplest and the most effective means of coping with the variation in soil hetrogeneity is to have a proper cholce of plots and blocks.

The experimental plot is the total amount of experimental material to which a treatment is applied in a single replicate: Any experimenter who wishes to conduct an experiment with any crop has to select a convenient plot size for conducting the experiment. In many situations a decision on the size and shape of plot is made arbitrarily depending solely up on the judgement and experience of the reaearch worker. But it is to be noted that an improper choice of the experimental unit (plot) can offset the experimental findings greatly. A very small plot even though appreciable from the economic point of view may give highly biased results. On the contrary, extremiy large plots result in mere wastage of resources at the cost of. very little gain in precision. Thus, it is always advantageous to use the most efficient plot size for conducting field trials. For a given size of plots, different geometerical configurations of the units are possible leading to various shapes of plot. Alternately, shape of.the plot can be defined by the ratio

L:B, where $L$ is the length of the plot and $B$ is the bregdth of the plot. It is also desirable to have an idea about the best shape of the plot which result in maximum precision for a given size of the plot. "Block" is a group of plots which are more or less homogeneous. Efficiency of blocking depends on the uniformity of plots within.: the block and heterogenelty between blocks. The investigator must know the best criterion for grouping or blocking the units Inorder to achieve maximum precision. For a given size of the plot the efficiency of local control depends largely on the size and shape of blocks. An extremaly large block can be as inefficient in error control as there was no blocking. The orientation of plots and blocks in a field is usually determined on the basis of the direction of the fertility gradient. A fertility contour map of the field is very helpful in this respect.

The statistical considerations governing the choice of suitable dimensions of the plot are the effect of size and shape of experimental units on the magnitude of error variance and consequent precision of treatment comparison and on the total cost of experimentation. Theoretically the best size and shape of plot is one which should give minimum variability in the estimate of population mean. This concept is refertes to as themiconstrained optimisation of the variability function. Looking on the same problem
from another angle of vision, the best size of the plot is the one which gives maximum information per unit cost. For an experimenter with limited resources, these two approaches will not be appealing. He may be interested in finding the optimum plot size in the sense that it gives estimates with pre-assigned degree of precision utilizing only the minimum amount of experimental material.

Replication or blocks should be so set up to control as much of the variation as possible resulting in the smallest experimental error variance. If a knowledge of the soil heterogeneity of the field is available, it could be utilised in setting up the blocks. Size of the block for a given design is determined by the number of treatments and the size of the plots. The upper limit on the replicate size depends largely on the character studied and nature of variability. Since an increase in block size is follwed by a consequent enhancement of error it is not desirable to have too much entries in a block. This problem can be delt with by using incomplete block designs. However, a crtical study of the experimental material alone will help the experimenter to formulate appropriate criteria for determining the size of the replicate.

Turmeric, curcuma longa. ㄴ. belongs to natural order Scitaminae and family Zingiberaceae to which the familiar ginger and cardamom also belong. In India, it is mainly
used as spices and in medicines. But, in foreign countries, it is well known for its curcumin content and is used as a natural colouring material especially for colouring food products and costly textiles. It has a good export value and is a regular foreign exchange earner for the country. The estimated world production of turmeric is around 1.6 lakh tonnes, of which India's contribution is about 1.5 lakhs tonnes. In India 92 per cent of the produce is consumed within the country and remaining is exported to foreign countries. The foreign exchange earning by turmeric ranks founth among the spices first three places being occupied by black peper, cardemom and ginger respectively. In India, turmeric is cultivated in an area of about 77, 400 ha . of this 433 ha (5.6\%) is in Kerala State. Kerala contributes to about 15.1 per cent of India's total export on turmeric and earns arround 76 lakhs of rupees annually. Besides Kerala, the states of Andhra Pradesh, Tamil Nadu, Bihar and Orissa are the other important turmeric producing states of India. The contribution of turmeric by Kerala works out only to 2.8 percent of that of India. The quality of turmeric expressed as curcumin content is very important in export market. But most of the Indian turmeric types contain less than 5 percent curcumin. The foregoing details stress the importance of turmeric in Indian economy., Therefore it is essential to conduct research with the objectives of improving


#### Abstract

the quality and yield of tusmeric. Information on the statistical designing of field experiments on turmeric is rather scanty. Field trials on turmeric are usually conducted by using the same size and shape of the plot required for ginger - a similar crop. Thus, there is an urgent necessity to have a deep Investigation in the field plot technique of experimentation exclusively on turmeric.


The present study undertaken on turmeric has the following objectives :-
(1) To study the nature and magnitude of soil heterogeneity of the experimental field.
(ii) To determine the optimum size and shape of plots for conducting field experiments on turmeric under normal field conditions.
(ii1) To determine the maximum number of plots of a given size which can be accommodated in a single block without confounding.
(iv) To determine the direction of the blocks to increase the efficiency of field experiments.
(v) To compare the estimates of optimum plot sizes obtained through different criteria of estimation.
(vi) To estimate the relative efficiency of alternate designs in laying out field trials.
(vii) To seek for alternate modelsfor describing the
relationship between plot size and variability.
(viii) To determine minimum number of replications required for estimating treatments effects with given degree of accuracy.

## REVIEW OF LITERATURE

In this chapter an attempt has been made to give an account of the research information on the technique of field experimentation of different crops.
2.1 . Magnitude of Soil Heterogeneity

Harris (1920) initiated studies on the statistical treatment of soil heterogeneity and its relation to the accuracy of experimental results. Through the estimation of intraclass correlation coefficient, he concluded that soil heterogeneity is the most potent source of variation in plot yields and the chief difficulty in their interpretations. He showed that the correlation between the Yields of adjacent plots was either due to initial physical and chemical similarities of the soil or to the influence of previous crops upon the nature and composition of the soil. The Intraclass correlation coefficienc of Harris (1915) served only to demonstrate the degree of difference in soil heterogeneity of adjacent plots. But Bose (1935) found that an experimental site which was reasonably uniform for one crop in one season was not necessarily uniform for another crop in another season. He concluded that analysis of variance was more useful than the intraclass correlation coefficient of Harris, because it provided
the natire of soil heterogeneity and permitted the identification of fertility gradients.

Smith (1938) proposed a quantitative measure of soil heterogeneity based on his empirical relationship between plot size and variability of mean per plot given by the equation,

$$
v_{x}=v_{1} x^{-b}
$$

Where $V_{x}$ is the variance of mean yield per plot based on plots of $x$ unit in size, $V_{1}$ is the variance among plots of size unity and $b$ is the index of soil heterogeneity, Which assumed values only in the range between zero and one. A value of 'b' nearer to one indicated that there was no significant correlation among contigöus units, whereas a value in the neighbourhood of zero indicated strong linear relationship between adjacent units. In the case of selffertilised crops the value ' $b$ ' was largely a function of the effect of soil heterogeneity, but with cross-fertilized crops intra-plot variation mainly due to genetic make up of the plants with plot also had some effect on the value of 'b'. A high value of 'b' tending to one thus indicated that genetic variation (intra-plot variation) was more predominant over positional variation. From a uniformity trial on cashew, Nair (1981) obtained the value 'b' as high as 0.97 whereas on oat, Handa et al (1982) obtained the values within the range 0.084 to 0.187 .

Federer (1955) observed that in most cases the value of heterogeneity coefficient calculated from the Smith's equation were in the range 0.3 to 0.7 . He further remarked that a change in plot size from one-fourth to four times the optimum will not greatly affect the cost or variance of heterogeneity in the normal range of 0.3 to 4. 2.2. Uniformity Trials and Fertility Contour Maps

An overall idea about the magnitude and distribution of soil heterogeneity of the experimental field can be obtained by conducting an experiment called "uniformity trial" which consists of growing a bulk crop with a uniform treatment all over the fleld and harvesting and recording the produce in small units of suitable size (Panse, 1941).

Cochran (1937) had given an account of 191 uniformity trials conducted on various crops by several workers. He noticed considerable variability in the estimates obtained from different crops and for the same crop in different locations.

Numerous reports on uniformity trials on various crops are available in India and abroad. To cite a few are those conducted by Baker et al (1952) on barley; Gopini et al (1970) on ground nut; Prabhakaran and Thomas (1974) on tapioca; Katyal and Sasmal (1982) on fute and Binns et al (1983) on tobacco.

Uniformity trial data can be presented graphically in what are called fertility contour map showing lines passing through areas of equal fertility. Fertility contour maps for numerous crops have been published by various workers. Some of them are those published by Hutchinson and Panse (1935a) on , cotton: Kadam and Patel (1937) on bajari; Agarwal et al (1968) on arecanut; and Hariharan (1981) on brinijal.

Several methods have been suggested from time to time by various workers for the estimation of convenient plot size for the conduct of successful field experiments. A brief account of the various methods' of estimation of optimum plot size is given below.

### 2.3.1. Maximum Curvature Method

Maximum curvature method consists in representing the relationship between plot size and coefficient of variation graphically by using a smooth free-hand curve and choosing the size of the plot just beyond the point of maximum curvature as the optimun (Federer, 1955). He has pointed out two weaknesses of this metiod. They are (i) the relative costs of various plot sizes are not considered and (if) the point of maximum curvature is not independent of the smallest unit selected or the scale of measurement used. In spite of
its inherent drawbacks several workers have used it for getting a suitable plot size due to its simplicity. Prabhakaran and Thomas (1974) used this technique for getting an initial crude estimate of plot size for field experiments on tapioca and Hariharan (1981) used it for estimating the plot size for field trials on brinjal.

### 2.3.2. Heterogeneity Index Method

Smith (1938) proposed a method for determining the optimum plot size from uniformity trial data. Smith's equation is given by $V_{x}=V_{1} x^{m b}$. Since the cost of experdmentation is also to be considered in determining a suitable plot size, he used the cost function of the form $K=K_{1}+K_{2} x$ where $\mathrm{K}_{1}$ is the cost assoclated with number of plots and $\mathrm{K}_{2}$ the cost associated with a unit area within the plot and $x$ the number of basic units per plot. The estimate of optimum plot size $x_{o p t}$, as suggested by Smith (1938) was given by, $x_{o p t}=b k_{1} /(1-b) k_{2}$.

Smith's equation has also been used by several workers to describe the non-linear relation between size of the plot and coefficient of variation (CV). Smith's equation In the modified form is given by $y=a x^{-b}$ where $y$ is the coefficient of variation per plot based on plots of $x$ units in size a is the coefficient of variation of plots of size unity and $b^{\prime}$ an index of soil heterogeneity related to Smith's ${ }^{\mathbf{t}} \mathrm{b}^{\prime}$ 。

Smith's equation in the modified form was used by Saxana et al (1972) on riot; ; Prabhakaran and Thomas (1974) on tapiocas and Hariharan (1981) on brinjal for estimating optimum plot size.

Raghavarao (1983) suggested that the optimum plot size could be determined from Smith's law in the modified form mathematically using calculus method by maximising curvature of the variability function. He estimated the optimum plot size of Radish using the new technique as 4 to 8 square meters.

Smith (1938) had not specifically defined the basis for calculating the factors $K_{1}$ and $K_{2}$ in the cost function. Marani (1963) pointed out that Smith's cost concept had been misused by several workers and indicated that the two types of costs should be proportional to $K_{1}$ and $K_{2} x$ and not to $\mathrm{K}_{1}$ and $\mathrm{K}_{2}$.

The correct definition of the cost functions were used for estimating optimum plot size by Hodnett (1953) in groundnut; Sen (1963a) in tea; Sreenath (1973) in sorghum; Prabhakaran and Thomas (1974) in tapioca; Biswas et al (1982) in cabbage and Binns et al (1983) in tobacco.

Hatheway and Williams (1958) presented a method of weighting of observed variances of plots of different sizes for getting an unbiased estimate of 'b' with asymptotically minimum variance.

### 2.3.3. Hatheway's Method

Hatheway (1961) developed a procedure, to determine optimum plot size, where the number of replication and the expected magnitude of difference betwen treatments were specified, but no attention was given to experimental cost. He used the relationship between coefficient of variation and Smith's 'b' in estimating plot size. The basic equation of Hatheway is of the form $x^{b}=2\left(t_{1}+t_{2}\right)^{2} c_{x}^{2} / r d^{2}$ where $x$ is the plot size, 'b' is an index of soil heterogeneity, $t_{1}$ is the observed value of $t$ in the test of significance, $t_{2}$ is the tabulated value of $t$ corresponding to $2(1-\mathrm{p})$ where $p$ is the probability of obtaining a significant result. $C_{x}$ is the coefficient of variation of plots of size $x$ units d is the true difference to be detected between two means expressed as a per centage and $I$ is the number of replications. He developed a set of curves for a specific set of conditions from which an experimentercan determine the proper plot size and number of replication for specified value of 'd'
2.3.4. Method of Estimation of Plot Size for Perennial Crops

Freeman (1963) suggested a modification to Smith's
law to take care of genetic variation among trees of the same plot. His new function is of the form
$\frac{v_{x}}{V_{1}}=-\frac{\alpha}{x^{b}}+\frac{(1-\alpha)}{x}$ where $v_{x}$ is the total variance of mean
yieid per tree of a plot containing $x$ trees, $V_{1}$ is the variance of the single tree plots, $\alpha$ is the proportion of the variance that is due to enviornment, $x$ is the number of basic units (trees) per plot and 'b' is the Smith's index of soil heterogeneity. Putting $\alpha=1$ in this equation, we get the familiar Smith's equation (1938). Freeman (1963) has also described'the method of estimating $\alpha$ by using serial correlations.

### 2.3.5. Method of Modified Maximum Curvature

Situations may often arise where the familiar Smith's law (1938) fails to describe the pattern of variability satisfactorily. Then either a change of scale or the need of fitting other sophisticated models is indicated. Lessman and Atkins (1963a) found that the equation $\log C_{x}=\frac{a}{(a+\log x)^{b}}$. where $C_{x}$ is coefficient of variation of plots of size of $x$ units, is more efficient in representing the relationship between plot size and varlability than Smith's law.

Prabhakaran (1983) suggested three non-linear models for describing the relationship between coefficient of variation and plot size ( $x$ ). He has shown enpirically that all these three models were superior to Smith's law in describing the proposed relationship between plot size and coefficient of variation at least for three different crops viz. tapioca, banana and cashew.

The suggested models are

$$
\begin{align*}
& Y=a+b / \sqrt{x}+c / x  \tag{i}\\
& Y^{-1}=a+b \log x \\
& Y^{-1}=a+b \sqrt{x}+c x
\end{align*}
$$

2.3.6. Varlance, Component Heterogeneity Index Method.

Koch and Rigney (1951) developed a new method called Variance Component Heterogeneity Index Method for estimating plot size by utilizing data from actual field experiments with different treatments and not from uniformity trial data. This method consisted in estimating the components of variance due to plots of different sizes by reconstructing the ANOVA of the specified design and using these estmated variances for fitting the Smith's'functions.

But Hatheway and Williams (1958) pointed out that the method of Koch and Rigney (1951) often resulted in inaccurate estimates of plot size because they assigned equal weights to the different components of variation even though they were based on different degrees of freedom.

### 2.3.7. Percentage Relative Efficiency Concept

Another approach in estimating plot size is to select the plot, size which gives maximum precision for given cost as optimum. If the rectprocal of the coefficient of variation
can be considered to be an index of precision, the efficiency of a plot can be defined as $1 / x C_{x}$, where $C_{x}$ is the coefficient of variation of plot size. $x$ estimated from the mathematice" model (Kalamkar, 1932). Therefore relative efficiency of plot size $\cdot x_{2}$ as compared with plot size $x_{1}$ is given by

$$
\mathrm{RE}_{12}=\frac{\mathrm{x}_{1} \mathrm{C}_{\mathrm{x} 1}}{\mathrm{x}_{2} \mathrm{C}_{\mathrm{x}_{2}}} \times 100
$$

Gopini et al (1970) had shown that efficiency of a plot decreased with an increase in size of the plot in the case of groundnut. Similar results were obtained by Sexana et al (1972) on oat; Sreenath (1973) on sorghum; Prabhakaran and Thomas (1974) on tapioca; Rambabu et al (1980) on fodder grass and Hariheran (1981) on brinjal.

Optimum plot size can be obtained by maximising information per unit area. It has been showed by various workers such as Menon and Tyagi (1971) on mandarin orange: BAُarghava and Sardana (1973) on apple and Prabhakaran et al (1978) on banana that single tree or plant plots were the most efficient ones for conducting field trials on these crops as they provided maximum amount of relative information.

It is to be noticed that both these approaches are identical and produce identical results.

The third approach of estimating the plot size is to
select the size of the plot which required minimum experimental material for a given preciston (Gomez, 1972).
2.4. Shape of plots

Taylor (1907-09) who summarised a large number of contemporary field experiments with various crops found that rectangular plots were the most desirable and convenient for experimentation with field crops.

The first theoretical consideration on the shape of the plot was given by Chivistidis (1931). He derived an expression for estimating the effect of plot shape on variability with the help of the assumption of a linear fertility gradient and concluded that long and narrow plots. are always more efficient than square ones. Many research workers agreed with his findings. They include, Saxana et al (1972) on oat; Sreenath (1973) on sorghum, Prabhakaran and Thomas (1974) on tapioca and Hartharan (1981) on bringal.

Cochran (1940) also considered variations in the shape of the plots for various types of field experiments. He attributed the cause of variation with small and large bands of fertility gradients present in the experimental field. He found that the selected plot shape did not exertal considerable effect on soil heterogeneity when the variation in fertility gradient was small whereas if there is significant variation in fertility pattern, long and narrow plots was found to alve
a better control of error variance than square plot.

Marcer and Hall (1911) working with mangoes found no superiority of long and narrow plots over square ones. Bist et al (1975) on potato found that shape of the plot had no consistent effect on estimates of error. Similar results have been reported by Rambabu et al (1980) on fodder grass and Biswas et al (1982) on cabbage.

Pan (1930) obtained contradictory results about plot shape of rice in China. At Hangehow increasing plot width was more efficient than increasing length whereas at Wufe the opposite was true.
2.5.

Size and Shape of Blocks

Panse (1941) considered the effect of size and shape of blocks and their arrangement on the magnitude of soil variation. He developed a concept of block efficiency for computing the relative efficiency of blocks of different sizes and shapes with regard to the power of sorting out the assignable component of variation due to difference among blocks from experimental error. He concluded,however, that size and shape of plots exerted greater influences on error variation than that of the blocks of a given experimental fleld and hence greater attention to be given on the appropriace choice of plots than that on blocks.

Iyer and Agarwal (19.70) found that compact blocks are more efficient than rectangular blocks in laying out experiments on sugarcane. Similar results were also obtained by Saxana et al (1972) on oat and Handa et al (1982) on oat.

Sreenath (1y/3), found tnat snape or tre Dlocks na no. consistent effect on block efficiency on sorghum and this result was supported by Bist et al (1975) on potato and Rambabu et al (1980) on fodder grass.

Gopini et al (1970) found that block efficiency decreased for given size and shape of plots with increase in the block size in groundnut and the result has been supported by the findings of Saxana et al (1972) on oat: Kripashankar et al (1972) on soyabean; Sreenath (1973) on sorghum, Bist et al (1975) on potato and Hariharan (1981) on brinjal.

### 2.6. Minimum Number of Replications

Hayes (1925) proposed the formula $r=C v^{2} / p^{2}$ for determining the minimum number of replications ( $r$ ), needed to estimate population mean with 'p' percent standard error; where Cv is the coefficient of variation. He showed that an increase in number of replications decreased standard error more rapidly than an increase in the size of the plot. Many research workers on various crops experienced the same
result. They include Iyer and Agarwal (1970) on sugarcane; Kripashankar et al (1972) on soyabean; Bist et al (1975) on potato: Prabhakaran et al (1978) on banana; Hariharan (1981) on.brinjal and Suman and Wahi (1982) on cabbage.

According to Gomez (1972) one of the simplest means of increasing the prectsion of treatment comparison is to increase the number of replications for different treatments but beyond a certain level, the improvement in precision attainable through the increase in number of replication is too small to worth the additional cost, other means of enhancing prectision have to be employed.

Prabhakaran et al (1978) on banana observed that the expected number of replications decreased with an increase in plot size but total number of experimental trees (plots) Increased with an increase in plot size.
2.7.

Relative Efficiency of Different Designs

Fisher (1951) had used the concept of relative efficiency for the choice of appropriate designs for conducting field trials. According to him relative efficiency is the ratio of amount of information supplied by one design to the amount of information supplied by another design.

Malhotra et al (1979) found that the relative efficiency
of latin square designs of different orders for different plot sizes compared with completely randomised designs ranged from 122 to 262 percentage whereas those when compared with randomised block design using either row or column as blocks ranged from 100 to 292 percentage. "Jayaraman (1979) found that the efficiency of randomised blocks design over completely randomised design'depended largely on the orientation of blocksp and that of latin square design'over randomised block design also depended on the orientation of the blocks of randoraised block design. He found that on an average the relative efficiencies of latin square design over randomised block design was 107.8 and 238.5 percentage for row as blocks and column as blocks respectively and for combined analysis (Federer, 1955) eliminating the between set sum of squares the efficiencies are 118.7 and 222.4 percentage for row as blocks and column as blocks respectively.

## MATERIALS AND METHODS

The study was taken up at the College of Horticulture, Vellanikkara during the period from June 1983 to January 1984 by conducting an unfformity trial on turmeric. The weather and seasonal conditions during the period of study were more or less normal. The experimental field selected was uniform with non-undulating topography and no shade trees and pathways around the margin. The soil was red lateretic loam. Adequate drainage was provided. The variety of turmeric under study was "Wyanad Local".

The crop was raised in raised beds adopting manurial and cultural operations as per package of practice recomended by Kerala Agricultural University.

The gross experimental area consisted of a rectangular field with sides of 74.2 meter length and 15.2 meters breadth. Small elevated beds of height of 0.25 cm and of size 0.6 m x 1.5 m were raised providing channels of width 0.4 m zaround each to prevent soil erosion and water logging. There were altogether 494 beds in the field. One row of beds all around the margin of the field was discarded to eliminate external border effect. After discarding the border rows, there were 432 beds in the net experimental area.

Harvesting was done on 220 th day after planting,

When the leaves had dried completely in most of the plants. At the time of harvest each bed was divided into equal plots of size $0.6 \mathrm{~m} \times 0.75 \mathrm{~m}$ and the yield was recorded from each plot for statistical analysis.
3.1. Fertility Contour Map

In order to construct fertility contour map, the percentage deviation of each observation from the grand mean was calculated by the relation,

$$
\begin{equation*}
d_{i}=\frac{\left(Y_{i}-\bar{Y}\right)}{\bar{Y}} \times 100 \tag{1}
\end{equation*}
$$

where $d_{i}=$ Percentage deviation of the ith unit from the grand mean
$Y_{1}=$ Yield on the ith unit
$\overline{\mathbf{y}}=$ Grand mean

The units are then grouped into different classes according to the magnitude of the observed variation saround the overall mean yield. The experimental units which produced the same amount of deviation from the overall mean yield was assumed to be similar in fertility. Regions of similar fertility status were identified and marked with different system of grading.

## 3.2. Size and Shape of Plots

Plots of different sizes and shapes were formed by grouping adjacent units in various possible ways. The mean, standard deviation and coefficient of variation of plots of different sizes and shapes were worked out. Optimum size and shape of plots were determined by using several methods as indicated below.

### 3.2.1. Maximum Curvature Method

A freehand, curve was drawn by joining the points ploted with absscisae equal to the sizes of the plot and ordinate equal to the corresponding coefficients of variation. The optimum plot size was determined from the curve as the one just beyond the point of maximum curvature.
3.2.2. Heterogeneity Index Method

Smith's (1938) empirical law is given by,

$$
\begin{equation*}
v_{x}=\frac{v_{1}}{x} e^{u^{\prime}} \tag{2}
\end{equation*}
$$

where $\mathrm{V}_{\mathrm{x}}=$ the variance of the yield per unit area among plots of $x$ unit in size.
$V_{1}=$ the variance among plot of one unit in size.
$x=$ the number of basic units in a plot.
$\mathrm{b}=$ the index of soil heterogeneity.
$e^{u}=$ random error component, where $u^{\prime}$ is $N\left(0, \sigma^{2}\right)$

Smith's empirical law expressed in the modified form is given by

$$
\begin{equation*}
\ddot{y}=a x^{-b} \tag{7}
\end{equation*}
$$

where $Y=$ expected coefficient of variation of the Yield per unit area among plots of $x$ units in size. $a=$ the coefficlent of varlation among plots of one unit in size:
$b^{\prime}=$ index of soil heterogeneity.

It is evident that,

$$
\begin{equation*}
\hat{y}^{2}=\frac{\mathrm{V}}{\mathrm{~m}^{2}} \tag{4}
\end{equation*}
$$

where M is the grand mean per unit basis.
and

$$
\begin{equation*}
\mathrm{a}^{2}=\frac{\hat{\mathrm{V}}_{1}}{\mathrm{M}^{2}} \tag{5}
\end{equation*}
$$

Substitute (4) and (5) in square of (3),

$$
\begin{equation*}
\hat{v}_{x}=\hat{v}_{1} x^{-2 b^{\prime}} \tag{6}
\end{equation*}
$$

Therefore from (2) and (6).

$$
b=2 b^{\prime} \text {. Smith's index of soil }
$$

heterogeneity:

The coefficient ' $b$ ' and the constant 'a' are obtained from the mathematical model

$$
\begin{equation*}
y=a x^{-b^{\prime}} e^{u} \tag{7}
\end{equation*}
$$

where 'u' is distributed as independent $N\left(0, \sigma^{2}\right)$

Taking logarithms of (7)
$\log Y=\log a-b^{\prime} \log x+u$
ie.

$$
\begin{equation*}
\underline{Y}_{1}=A+\underset{\sim}{B} X_{1}+u \tag{9}
\end{equation*}
$$

where

$$
\begin{aligned}
& \mathrm{X}_{1}=\log Y \\
& \underset{\sim}{A}=\log A \\
& \text { and } \quad=-b^{\prime} \\
& \mathrm{X}_{1}=\log x
\end{aligned}
$$

By the method of Ieast squares $A$ and $B$ can be obtained from the normal equations
where'

$$
\begin{equation*}
X_{1} Y_{1}=\left(X_{1} X_{1}\right) A_{1} \tag{10}
\end{equation*}
$$

From this normal equation, the coefficient b' and constant 'a' are obtained.

For calculating an optimum plot size consider the cost function of the form,

$$
K=k_{1}+k_{2} x
$$



The optimum plot size is obtained by minimising total cost per unit of information. That is, by minimising $C$. where

$$
c \quad=\quad\left(k_{1}+k_{2} x\right) / \frac{1}{V_{x}}
$$

ie.

$$
\begin{equation*}
c \quad=\quad a^{2} M^{2}\left(K_{1}+K_{2} x\right) / x^{b} \tag{11}
\end{equation*}
$$

On differentiating $C$ with respect to $x$ and equating to zero the optimum size of the plot which give maximum information per unit cost is obtained.
ie.

$$
\frac{d c}{d x}=0
$$

or $\frac{d \log C}{d x}=0$ gives

$$
\frac{K_{2}}{K_{1}+K_{2} x}=\frac{b}{x}
$$

$$
\begin{equation*}
\because \quad \hat{x}=\frac{b k_{1}}{(1-b) k_{2}} \tag{12}
\end{equation*}
$$

put $b=2 b^{\prime}$ in (12)

$$
\begin{equation*}
\hat{x}=\frac{2 b^{1} k_{1}}{\left(1-2 b^{i}\right) k_{2}} \tag{13}
\end{equation*}
$$

and $C$ is minimum at $\dot{x}=\hat{x}$ only if $\frac{d^{2} \log c}{d x^{2}}$ at $x=\hat{x}$ is greater than zero.

Optimum plot sizes for different cost ratios can be determined from the formula (13) by assigning different values for the cost components $K_{1}$ and $K_{2}$.
3.2.3. Modified Maximum Curvature Method using Smith's Equation

The curvature $C$ of a line at a given point is defined as the limit of the average curvature of the arc, when the length of the arc approaches zero. Average curvature means $\varphi / \widehat{A B}$, where $\widehat{A B}$ is the arc and $\varphi$ is the angle formed by the tangents at $A$ and $B$. That is, by definition $C=\frac{d \varphi}{d s}$, where $s=A B$

$$
\therefore \quad c=\frac{d \varphi / \partial x}{d s / \partial x}
$$

let $Y=f(x)$, the function of $x$, then

$$
\begin{equation*}
\tan \varphi=\frac{d y}{d x} \tag{14}
\end{equation*}
$$

Defferentiating (14) with respect to $x$

$$
\begin{equation*}
\frac{d \varphi}{d x}=\frac{d^{2} y}{d x^{2}} / 1+\left(\frac{d y}{d x}\right)^{2} \tag{15}
\end{equation*}
$$

IE Is $=\underset{B \rightarrow A}{ } \quad \widehat{A B}$

$$
d s=\sqrt{d y^{2}+d x^{2}}
$$

$\cdot \frac{d s}{d x}=\sqrt{1+\left(\frac{d y}{d x}\right)^{2}}$

Put $\frac{d y}{d x}=y_{1}$ and $\frac{d^{2} y}{d x^{2}}=y_{2}$, then

$$
\begin{equation*}
c=\frac{Y_{2}}{\left(1+Y_{1}^{2}\right)^{3 / 2}} \tag{16}
\end{equation*}
$$

The optimum plot size is the point at which the curvature is maximusn. That is, $C$ is maximum. The point of
maximum curvature is obtained by equating

$$
\frac{d C}{d x}=0 \quad \text { or } \quad \frac{d \log C}{d x}=0
$$

Smith's empirical law in the modified form is

$$
y=a x^{-b} .
$$

Q. $\quad Y_{1}=-a b^{\prime} x^{-\left(b^{\prime}+1\right)}$ and $Y_{2}=a b^{\prime}\left(b^{\prime}+1\right) x^{-\left(b^{\prime}+2\right)}$

$$
\text { ie. } \quad c=\frac{a b^{\prime}\left(b^{\prime}+1\right) x^{-\left(b^{\prime}+2\right)}}{\left(1+\left(a b^{\prime}\right)^{2} x^{-2\left(b^{\prime}+1\right)}\right) 3 / 2}
$$

$\therefore \frac{d \log C}{d x}=\frac{-\left(b^{\prime}+2\right)}{x}+\frac{3\left(a b^{\prime}\right)^{2}\left(b^{\prime}+1\right) x^{-\left(2 b^{\prime}+3\right)}}{1+\left(a b^{\prime}\right)^{2} x^{-2\left(b^{\prime}+1\right)}}$

Equating to zero, gives

$$
\begin{aligned}
& x^{2\left(b^{4}+1\right)}=\frac{\left(a b^{\prime}\right)^{2}\left(2 b^{1}+1\right)}{\left(b^{2}+2\right)} \\
& \text { If this value of } x \text { is substituted in } \frac{a^{2} \log C}{a x^{2}} \text { it will }
\end{aligned}
$$ be less than zero. .Then the optimum plot size can be determined from the equation (17).

3.2.4. Alternate Models

Three other non-linear models were also tried to express the relation between coefficient of variation and plot sizes.

The three models are,
(1) $\quad Y=a+b / \sqrt{x}+c / x+u$ $y^{-1}=a+b \log x+u$
(iii) $Y^{-1}=a+b \sqrt{X}+c x+u$

In all the three models the parameters were estimated by principles of least squares.

The mathematical method by using calculus to find the optimum plot size is still applicable to these models also. 3.3. Relative Efficiency of Plot Sizes

Kalamkar (1932) defined efficiency of a plot of size. $x$ units as $1 / x C x$, where $C_{x}$ is the coefficient of variation of a plot of size $x$ unit. The relative efficiency of plot of size $P_{2}$ as compared with a plot of size $P_{1}$ is defined as the ratio of the efficiency of $P_{2}$ over that of $P_{1}$ and is denoted by $R E\left(P_{2} / P_{1}\right)$.

Thus if $x_{1}$ and $C_{x 1}$ are the number of basic units and coefficient of variation of a plot of size $p_{1}$ and $x_{2}$ and $C_{x 2}$
are the number of basic unit and coefficient of variation of a plot of size $P_{2}$, then

$$
\begin{equation*}
\operatorname{RE}\left(P_{2} / P_{1}\right)=\frac{x_{1} C_{x 1}}{x_{2} C_{x 2}} \tag{18}
\end{equation*}
$$

3.4.

Relative Efficiency of Blocks

The advantage of uaing blocks of different sizes in reducing experimental error by removing a portion of variability due to them is called block efficiency: This can be measured by "finding the inverse of error mean square obtained after the elimination of difference due to blocks of specified size from the total variation.

The relative efficiency of a block of size $P_{2}$ when compared with a block of size $p_{1}$ is defined as the ratio of the efficiency of block of size $P_{2}$ over that of $P_{1}$. This can be expressed in percentages.

Smith (1938) deduced the following relationship for the variance per unit area between plots ${ }_{\kappa} x$ units in blocks of $m$ plots. $V_{x m}=\frac{m\left(1-m^{-b}\right)}{m-1} V_{x^{*}}$ where $V_{x}$ is the variance in an Infiniteland and $b$ is the Smith's index of soil heterogeneity.

Therefore the efficiency of blocks of m plots relative to blocks of $n$ plots is equal to

$$
\operatorname{RE}(m / n)=\frac{V_{x n}}{V_{x m}}
$$

ie.

$$
\begin{equation*}
\operatorname{RE}(m / n)=\frac{n(m-1)\left(1-n^{-b}\right)}{m(n-1)\left(1-m^{-b}\right)} \tag{19}
\end{equation*}
$$

This concept has also been used for calculating the efficiency of different block sizes.
3.5. Number of Replications and Area Required

The minimum number of replication for estimating means with $P \%$ standard error was worked out for different sizes of plots and blocks by using the formula

$$
\begin{equation*}
r=C v^{2} / D^{2} \tag{20}
\end{equation*}
$$

where $r=$ minimum number of replication
$C V=$ estimated coefficient of variation
and $\quad P=$ the percentage standard error of the mean

The total area required for experimentation can be obtained by multiplying the area of the plot with the number of replication at $P \%$ standard error of the mean.

Assuming a simple cost function of the form, $C=r k x$, where $r$ is the number of replication, $k$ is the cost per unit plot size and $x$ is the number of basic unit per plot. The size of the plot which requiresminimum experimental
material is also the best plot size in the sense that it results in a minimum experimental cost for a given degree of precision. The optimum plot size has also been estimated on the basis of this concept.
3.6. Relative Efficiency of Designs

Relative efficiency ( $R E$ ) of a design $D_{1}$ over another design $D_{2}$ is primarily defined as the ratio of the amount of information supplied by $D_{1}$ over $D_{2}$.

$$
\begin{equation*}
\text { ie. } \quad \operatorname{RE}\left(D_{1} / D_{2}\right)=\frac{1 / \operatorname{ci}_{2} 2}{1 / D_{2}} \tag{21}
\end{equation*}
$$

Where $\sigma_{i}^{2}=$ the expected value of error variance of experimental design $D_{1}$
and $\sigma_{2}^{2}=$ the expected value of error variance of experimental design $\mathrm{D}_{2}$. $\frac{2}{\sigma_{1}}$ and ${\frac{t_{i}}{2}}_{2}$ are estimated as $s_{1}{ }^{2}$ and $s_{2}^{2}$ with respective degrees of freedom $\mathrm{V}_{1}$ and $\mathrm{V}_{2}$. Then relative efficiency (RE) can be estimated by the formula suggested by Fisher (1951) as,

$$
\begin{equation*}
\operatorname{RE}\left(D_{1} / D_{2}\right)=\frac{s_{2}^{2}\left(v_{1}+1\right)\left(v_{2}+3\right)}{s_{1}^{2}\left(v_{2}+1\right)\left(v_{1}+3\right)} \tag{22}
\end{equation*}
$$

In this study the relative efficiency of three types of design alone were compared viz. completely Randomised

Design (CRD), Randomised Block Design (RBD) and Latin Square Design (LSD).

For the comparison of relative efficiencies plots of size 6 units with all possible shapes also were considered. In the case of plot: size' $1 \times 6$. nine laṭin squares of order $4 \times 4$ and one latin: square of order $12 \times 12$ can be formed whereas in the cases of $2 \times 3$ plota arrangementsfour latin squareSof order $6 \times 6$ and for'3 x 2 plot arrangements nine latin square of order $4 \times{ }^{4} \mathrm{n}$. The relative efficiency can be obtained in two ways for each dimension of the plot.
(i) The relative efficiency is first determined for each of the squares and average of these taken as the representative for the entire area using the formula in the modifled form as suggested by Federer (1955).

$$
\begin{equation*}
\operatorname{RE}(L S D / C R D)=\frac{\left(V_{1}+1\right)\left(V_{2}+3\right)(R+C+(k-1) E)}{\left(\mathrm{V}_{2}+1\right)\left(\mathrm{V}_{1}+3\right)(\mathrm{K}+1) \mathrm{E}} \tag{23}
\end{equation*}
$$

and $\quad \operatorname{RE}(L S D / R B D)=\frac{\left(V_{1}+1\right)\left(V_{2}+3\right)(R+(k-1) E)}{\left(V_{2}+1\right)\left(V_{1}+3\right) k E}$.

Where $\mathrm{R}=$ mean sum of squares of row
$\mathrm{C}=$ mean sum of squares of column
$E=$ mean square error in LSD
$k=$ the order of the LSD
and $V_{1}$ and $V_{2}$ as defined above
(ii) A combined analysis of sets of latin squares (Federer, 1955) also attempted and relative efficiency is determined after eliminating variation between sets of latin squares: Here,
$R E(L S D / C R D)=\frac{(s-1) S+s(k-1)(R+C)+s(k-1)^{2} E}{s\left(k^{2}-1\right) E} \times \frac{\left(V_{1}+1\right)\left(V_{2}+3\right)}{\left(V_{2}+1\right)\left(V_{1}+3\right)}$.
and $\operatorname{RE}(\mathrm{LSD} / \mathrm{RBD})=\frac{(\dot{s}-1) \mathrm{s}+\mathrm{s}(\mathrm{k}-1) \mathrm{C}+\mathrm{s}(\mathrm{k}-1)_{\mathrm{E}} \mathrm{E}}{(\mathrm{sk}(\mathrm{k}-1)+(\mathrm{s}-1)) \mathrm{E}} \times \frac{\left(\mathrm{V}_{1}+1\right)\left(\mathrm{V}_{2}+3\right)}{\left(\mathrm{V}_{2}+1\right)\left(\mathrm{V}_{1}+3\right)}$
for row as blocking.

It is hoped that this relative efficiency would reflect the overallrelative efficiencies of Latin Square Design over Completely Randomised Design and Latin Squaré Design over'Randomised Block Design, when the experiment involves sets of latin squaresin a single experiment.

RESULTS

The results of investigation carried out to estimate the optimum'size and shape of plots and blocks in turmeric are presented below.
4.1. Fertility Contour Map of the Experimental Field

The fertility contour map of the experimental field is given in Figure 1. An inspection of the fertility contour map indicated that there was appreciable variation in soil fertility but this variation did not follow any systematic pattern. Fertility variations were distributed over the entire field in an erratic fashion. It could also be seen that small areas were relatively more homogeneous with regarded to soil fertility than large areas.:
4.2. Estimation of Optimum Plot Size

Adjacent units were combined together to form plots of different sizes and shapes. Plot length was defined as the length in the North-South direction and plot breadth that in the East-West direction. The basic unit of observation consisted of the plant population in an area of size $0.6 \mathrm{~m} x$ 0.75 m . There were ten plants in the basic unit. The coefficients of variation (cv) for plots of different sizes and shapes when they are not grouped into blocks are given in

Table 1. It can be seen that CV decreased on either direction with an increase in plot size but the decrease was not proportional. Moreover reduction in cv in the North-South (column) direction was more rapid than that in the East-West (row) direction. Minimum cv noticed was aground 21 percent and the maxtmum was 77 percent. .
4.2.1. Method of Maximum Curvature

Smooth freehand curves were drawn (Fig.2) to represent the relationship between plot size $x$ and average ov when plots were not grouped in to blocks and when they were grouped into blocks of various sizes such as 2. 8, and 12 plots. It was found that in all the cases cv decreased rapidly at first when the size of the plot was increased, but after a certain point the rate of decrease was low and ultimately tended to zero making the curve almost like a straight inne parallel to the $x$-axis. The optimum plot sizes estimated from such freehand curves by the method of maximum curvature by visual inspection were 6 units for blocks of various sizes and 8 units without blocking. In original unit the optimum plot sizes for blocks of different sizes and without blocking were $2.7 \mathrm{~m}^{2}$ and $3.6 \mathrm{~m}^{2}$ respectively.
4.2.2. Smith's Equation in the Modified Form

The Smith's equations fitted to the uniformity trial
data on tumeric are given in Table 5a." The expected percentage variation, which could be explained by the fitted models was determined by calculating the coefficient of determination: $\left(R^{2}\right)$ and their significance tested through the variance ratio. (F)test. All the regression equations fitted to different blocks of stzes $2,4,8,12$ and 24 plots were found to besignificante Values of coefficient determination ranged from 0.8586 to 0.9883 . The parameters of the fitted models viz. ' $\mathrm{J}^{\prime}$ ' and ' $\mathrm{a}^{\prime}$ assumed value in the range 0.1223 to 0.1946 and 47.76 to 78.1088 respectively. 'Thus the Smith's index of soll heterogeneity $\left(b=2 b^{\prime}\right)$ varled between 0.2446 to 0.3882. Since the value of ' $b$ ' was nearer to zero than unity there appeared to be strong correlation between neighbouring plots. Hence proper orientation of plots and blocks is very inportant in controling experimental error. The expected cv (minimum co was dround 32, for block size 12 and the maximum was 78) was given in Table 3a. There was close agreement between observed and expected value of cv .
4.2.3. Other non-linear Models

Three other models viz.
and

$$
\begin{align*}
& Y=a+b / \sqrt{x}+c / x  \tag{i}\\
& Y^{-1}=a+b \log x \\
& Y^{-1}=a+b j \bar{x}+c x
\end{align*}
$$

Where Y is the coefficient of variation of plots of size $x$ units
were fitted to the data. The expected percentage of variation which could be explained by the fitted regressions (the coefficients of determination) and F-ratios resulted from the above models are given In Table $5 \& 6$. It could be seen that all these three models were more efficient than the familiar Smith's equation in describing the proposed relationship between plot size and coefficient of variation. Among the three models, Model 1 via, $Y=a+b / \sqrt{x}+c / x$ was found to be the best choice. It gave a very good fit to the uniformity trial data as the coefficient of determination calculated from that model was fairly high ( $\mathrm{R}^{2}$ ranging from 0.9403 to 0.9904 ). The expected cv determined from, this model is given in Table 3b. For Model 2, $\mathrm{R}^{2}$ varied from 0.8832 to 0.9842 and for Model 3, it varied between 0.8766 and 0.9765 .

### 4.2.4. Optimum Plot Sizes under Consideration of Cost

The optimum plot sizes calculated on the basis of arbitrary values for the ratio $k_{1}: k_{2}$ where $k_{1}$ is the overall cost of experimental unit of size $x, k_{2}$ is the cost of individual item within the experimental unit area are presented in trable 4. In calculating the optimum plot sizes, estimates; of ' $b$ ' from blocks of different sizes and without blocking and an average value of ' $b$ ' were used. When the ratio $k_{1} / k_{2}$ varied from $1 / 13$ to 13 the optimum plot
sizes varied from 0.0342 to 5.78863 units in two plot blocks, whereas the range of variation of blocks of sizes 4. 8, 12. 18 and without blocking were 0.037 to 6.253 , 0.0263 to $4.4408,0.0249$ to $4.2094,0.0281$ to 4.7450 and 0.049 to 8.2836 units respectively. For ${ }_{a}^{\text {the }} \mathbf{a v e r a g e}$ value of 'b' the optimum plot sizes varied from 0.0333 to. 5.6199 units $\left(0.015 \mathrm{~m}^{2}\right.$ to $2.53 \mathrm{~m}^{2}$ ) $\because$ It was found that the optimum size of the plot increased with an increase in the magnitude of cost ratio. If $k_{1}$ were the major contributor to the total cost than $k_{2}$ it would be more advantageous to use larger plots. In any case there was no significant advantage by using very large plots.

### 4.2.5. Optimum Plot Sizes using Smith's Modified Equation by Mathematical Methods

The optimum sizes of the plots for blocks of different sizes and without blocking determined by maximising curvature of the Smith's equation, $Y=a x^{-b}$, using the method of calculus are presented in Table 8. The estimated optimum plot sizes ranged between 4.443 units for blocks of size 8 unit and 8.1 units in case of no blocking. On an average the optimum plot size ${ }^{\text {is }}$ about 6 units ( $2.7 \mathrm{~m}^{2}$ ).

Concept of Percentage Relative Efficiency of Different Plot Sizes

Taking the efficiency of the smallest plot size as unity, the percentage relative efficiencies of various plot sizes are given in Table 7. As plot size increased the percentage relative efficiency decreased for blocks of different sizes and without blooking. When plots are not grouped the rate of decrease in efficiency was from 100 percent with unit plot size to 13 percent with plots of size 12 units. The percentage relative efficiency of plot of size 12 units compared with unit plot size for block of size 2,4,8,12 and 18 and without blocking were 12.46. 12.30, 11.54, 11.31, 11.88 and 13.28 respectively. Thus size of block had no significant effect in the efficiency of plots of different sizes. However, small sized blockswere more efficient.
4.4. Shape of the plot

For a given size of the plot the shape of the plot which gives least Cv may be selected for experimentation. The plot shape with least ov for different plot sizes and shapes are given in the Table 9. Thus for the optimum plot size viz. 6 units the optimum shape is $6 \times 1(3.6 \mathrm{~m} \times 0.75 \mathrm{~m})$ 。 In general. plot shape did not seem to exert any consistent effect on cv. However for a given plot size, long and narrow plots give lower cv then approximately square plots.
4.5.

Size and Shape of Blocks

The relative efficiencies of blocks of different sizes viz. 2,4,8,12 and 18 compared with no blocking calculated on the basis of percentage reduction in error sum of squares are given in Table 10. It can be seen that there was significant reduction in uncontrolled variation due to grouping of plots into blocks. Smaller the block size greater was the efficiency of blocking. In 2 plot blocks relative efficiency ranged from 207 to 262, whereas in 12 plot blocks it varied between 110.23 and 159.68 depending on the size of the plots. In general, size of the plot did not seem to exert any appreciable effect on block efficiency. Whereas for a given plot size, size of block had a significant effect on block efficiency. Thus for a plot of size unity relative efficiency decreased from 244.74 to 141.41 as the block size increased from 2 to 18. This fact indicated the need for reducing block size by way of using incomplete block designs.

Relative efficiency of blocks of different sizes as estimated from Smith's equation is given in Table 12. The entries in the second column of the table were calculated on the basis of assuming an average coefficient of heterogeneity (Smith's index of soil heterogeneity) $b=0.3018$ and those in the third column were estimated by assuming
the index of soil heterogeneity 0.3819 , which was the index of soil 'heterogenelty for Smith's function fitted whin for plots were not grouped. Block efficiency was found to decreasec as the size of the block increased. Four plot: blocks were almost $85 \%$ as efficient as 2 plot blocks whereas 6 plot: blockswere less efficient than 4 plots blocks. The relative efficiencies of blocks of size $4,6,8$ and 12 compared with blocks of size 2 were in the order 85, 78, 75 and 70 percentages when the index of soil heterogeneity for the ungrouped data was used. The corresponding figure with average index of soil heterogeneity were 83, 75, 70 and 60 percentages.

The shape of the block did not seem to exert any significant and consistent influence on the efficiency of blocking (Table 11) whereas block efficiency was found to be a function of the plot size. For 2 plot blocks of shape $2 \times 1$ the relative efficiency with plots of 2 units was 194.2 percent but that of 8 units was 298 percent. For 12 plot blocks in different shapes relative efficiency varied between 125.9 to 198.75 with plots of different sizes. No conslstent differences were observed between oblong blocks and compact blocks with regards: to their relative efficiencies in controlling error. However long and narrow blocks appear to be slightly more advantageous.

The minimum number of replication and the total experimental area required for estimating treatment means with 5 percent standard error are given in Table 13. From the table it could pe seen that with an increase in plot size, the expected number of replications decreased but the decrease was not proportional. The area required for the experiments also increased along with an increase in plot size. For example the numer of replication with a plot of size unity in blocks of size. 2 was 91 and that with a plot of size 12 was 42. Consequently the sizes of the experimental area required were $40.95 \mathrm{~m}^{2}$ and $226.8 \mathrm{~m}^{2}$ respectively. Thus if the slze of the field is fixed it was found to be more beneficial to use larger number of replications with the smallest possible plot size than increasing the plot size at the risk of reducing the number of replications. 4.7. Efficiency of Experimental Designs

Relative efficiency of a Latin Square Design (LSD) over the other two single factor designs viz. Completely Randomised Design (CRD) and Randomised Block Design (RBD) were calculated by using two methods (i) averaging, the relative efficiencies of different sets of squares and (ii) eliminating the variation between sets of squares and then calculating the relative efficiencies. The results are given in Table 14.

The average relative efficiencies of $6 \times 6$ Latin Square Design over Randomised Block Designs with rows as blocks and those with columns as blocks were 178.24 and 131.72 respectively, The relative efficiencies of the Latin Square of the same order by eliminating variation between squares compared with Randomised Block Designs with rows as blocks and columns as blocks were 267.33 and 147.05 respectively. The relative efficiencies of Latin Square Designs over Completely Randomised Designs are given in Table 1 fb It could be seen that Latin Square Designs were more efficient than Completely Randomised Designs in both the cases of eliminating variations among sets of squeres and averaging the relative efficiencies of different sets of squares. For the plot size $1 \times 6$ Randomised Block Designs with columns as blocks was found to be $84 \%$ more efficient than Completely Randomised Designs. Whereas for the same plot size Randomised Block Designs with rows as blocks was slightly more efficient (1.94\%) than Completely Randomised Designs. Thus, the relative efficiency of Randomised Block Designs depended upon the orientation of blocks.

TABLE 1. Observed Coefficients of variation for plots of different sizes when plots are not grouped

| Number o |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 1 | 2 | 3 | 4 | E | 8 | 9 | 12 | 18 | 24 | 36 | 72 |
| OU | 1 | 77.19 | 72.63 | 62.52 | 61.01 | 55.62 | 52.52 | 50.95 | 48.47 | 41.01 | 35.72 | 28.29 | 17.90 |
| $\begin{aligned} & 0 \% \\ & 0 . \\ & \text { rod } \end{aligned}$ | 2 | 64.92 | 61.91 | 55.11 | 54.02 | 50.24 | 47.33 | 46.27 | 43.99 | 37.25 | 31.33 | 24.71 | 13.62 |
|  | 3 | 58.60 | 56.83 | 51.67 | 50.94 | 47.07 | 45.22 | 42.75 | 41.57 | 33.73 | 29.45 | 21.91 | 11.17 |
|  | 4 | 56.38 | 54.97 | 50.65 | 49.97 | 46.90 | 45.19 | 43.56 | 42.07 | 35.33 | 29.24 | 21.32 | 13.45 |
|  | 6 | 48.68 | 47.63 | 44.92 | 44.01 | 42.78 | 41.18 | 39.51 | 38.53 | 32.17 | 26.24 | 21.32 | 12.08 |
|  | 12 | 43.55 | 43.13 | 41.50 | 41.08 | 40.90 | 39.62 | 38.02 | 38.13 | 32.95 | 26.83 | 23.15 | - |

TABLE 2. Observed Coefficient of Variation af plots of different sizes, when plots are grouped into different sizes and without grouping

| Plot <br> size (in units) | With- <br> out <br> block- <br> ing | Block sizes |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 2 | 3 | 4 | 6 | 8 | 9 | 12 | 18 | 24 | 36 | 72 |
| 1 | 77.19 | 49.34 | 58.64 | 56.27 | 60.39 | 59.32 | 61.35 | 62.82 | 64.91 | 65.63 | 67.70 | 72.58 |
| 2 | 68.89 | 46.86 | 52.27 | 50.83 | 53.59 | 52.47 | 55.74 | 57.34 | 58.45 | 58.51 | 62.62 | 64.84 |
| 3 | 60.59 | 38.58 | 40.99 | 42.03 | 45.59 | 45.80 | 46.09 | 48.95 | 50.34 | 53.04 | . 53.62 | 57.34 |
| 4 | 59.81 | . 38.71 | 43.18 | 41.63 | 46.28 | 42.90 | 48.38 | 47.33 | 51.97 | 51.06 | 56.73 | 57.64 |
| 6 | 54.16 | 33.81 | 37.76 | 38.17 | 40.76 | 40,84 | 41.46 | 46.97 | 46.46 | 48.85 | 50.45 | 52.22 |
| 8 | 53.85 | 33.24 | 39.65 | 34.76 | 39.02 | - | 46.53 | 45.95 | 49.37 | - | 51.55 | - |
| 9 | 51.31 | 35.60 | 35.40 | 38.47 | 40.10 | 41.87 | - | 45.72 | - | 47.80 | - | - |
| 12 | 48.65 | 30.44 | 34.87 | 34.06 | 38.33 | 38.17 | 40:67 | 42.51 | 44.61 | 44.56 | 47.25 | $\stackrel{-}{-}$ |
| 18 | 44.88 | 30.53 | 32.56 | 35.23 | 36.30 | 39.76 | - | 42.74 | - | 43.36 | - | - |
| 24 | 43.31 | 27.76 | 33.72 | 31.53 | 36.83 | - | 39.79 | 40.81 | 42.03 | - | - | - |
| 36 | 39.51 | 28.97 | 30.79 | 33.54 | 33.93 | 36.96 | - | 39.79 | - | - | - | - |
| 72 | 32.16 | 26.23 | 28.24 | 25.66 | 31.61 | - | - | - . | - | - | - | - |

TABLE 3a. Estimated Coefficients of Variation for plots of different sizes in blocks of different sizes and without blocking as obtained from the Smith's
equation, $Y=a x^{-b}$

| Block sizes | Plot sizes |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 4 | 6 | 8 | 12 |
| Without blocking | 78.101 | 68.253 | 59.640 | 55.115 | 52.115 | 48.161 |
| 2 | 47.764 | 42.928 | 38.582 | 36.246 | 34.676 | 32.577 |
| 4 | 53.444 | 47.754 | 42.670 | 39.951 | 38.127 | 35.698 |
| 8 | 54.954 | 50.313 | 46.064 | 43.746 | 42.173 | 40.052 |
| 12 | 59.457 | 54.624 | 50.184 | 47.756 | 46.105 | 43.875 |
| 18 | 62.561 | 57.023 | 51.976 | 49.234 | 47.376 | 44.876 |

TABLE 3b. Estimated Coeffcients of Variation for plots of different sizes in blocks of different sizes and without blocking as obtained from the model, $Y=a+b / \sqrt{x}+c / x$

| plot sizes |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Block sizes | 1 | 2 | 4 | 6 | 8 | 12 |
| Without blocking | 68.617 | 61.005 | 52.005 | 47.062 | 43.835 | 39.736 |
| 2 | 50.417 | 44.095 | 38.606 | 35.906 | 34.217 | 32.137 |
| 4 | 56.600 | 48.371 | 41.932 | 38.916 | 37.071 | 34.835 |
| 8 | 59.906 | 50.057 | 44.298 | 42.065 | 40.828 | 39.451 |
| 12 | 63.133 | 54.872 | 49.185 | 46.707 | 45.242 | 43.516 |
| 18 | 65.091 | 56.758 | 50.867 | 48.258 | 46.702 | 44.857 |

TABLE 4 Optimum plot sizes as' estimated from Smith's equation considering different cost ratios


Estimates of Parameters ; .. Coefficients of determinationRãa F-ratios from varlous models

TABLE 5a. Smith's equation, $y=a x^{-b}$

| Block sizes | a | b | $\mathrm{R}^{2}$ | F |
| :---: | :---: | :---: | :---: | :---: |
| Without blocking | 78.1088 | 0.1946 | 0.9883 | 843.04 |
| 2 | 47.7639 | 0.1540 | 0.9244 | 122.25 |
| 4 | 53.4441 | 0.1624 | 0.9127 | 10.457 |
| 8 | 54.9541 | 0.1273 | 0.8586 | 42.50 |
| 12 | 59.4566 | 0.1223 | 0.9183 | 101.14 |
| 18 | 62.5605 | 0.1337 | 0.9099 | 60.57 |
| 24 | 63.8999 | 0.1417 | 0.9723 | 210.75 |

TABLE 5b. Model $y=a+b / \sqrt{x}+c / x$

| Block sizes | a | b | $c$ | $R^{2}$ | $F$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Without |  |  |  |  |  |
| blocking | 17.9375 | 85.5918 | -34.9122 | 0.9827 | 255.39 |
| 2 | 21.8821 | 38.3626 | -9.8279 | 0.9518 | 88.93 |
| 4 | 24.2817 | 38.2772 | -5.9535 | 0.9403 | 70.85 |
| 8 | 34.5100 | 13.7545 | 11.6410 | 0.9664 | 78.90 |
| 12 | 35.9879 | 25.6430 | 1.5022 | 0.9520 | 79.31 |
| $18{ }^{\prime}$ | 36.6494 | 28.4288 | 0.0131 | 0.9440 | 42.11 |
| 24 | 35.9489 | 32.7503 | -2.9381 | 0.9904 | 257.31 |

Estimates of Parameters $-\cdots$ Coefficients of determinationk ${ }^{2}$ and F-ratios from various models fitted to the data with blocks of different sizes without blocking
TABLE 6a. Model $Y^{-1}=a+b \log x$

| Block sizes | a | b | $\mathrm{R}^{2}$ | $F$ |
| :---: | :---: | :---: | :---: | :---: |
| Without blocking | 0.0188 | 0.0099 | 0.9196 | 114.33 |
| 2 | 0.0206 | 0.0101 | 0.9413 | 160.38 |
| 4 | 0.0178 | 0.0099 | 0.9044 | 94.62 |
| 8 | 0.0180 | 0.0064 | 0.8832 | 52.94 |
| 12 | 0.0161 | 0.0058 | 0.9435 | 150.33 |
| 18 | 0.0159 | 0.0060 | 0.9243 | 73.26 |
| 24 | 0.0158 | 0.0062 | 0.9822 | 331.82 |

TABLE 6b. Model $\mathrm{Y}^{-1}=a+b \sqrt{\mathrm{X}}+\mathrm{cx}$

| Block sizes | a | b | c | $\mathrm{R}^{2}$ | F |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Without |  |  |  |  |  |
| blocking | 0.0085 | 0.0038 | -0.0004 | 0.9350 | 64.77 |
| 2 | 0.0157 | 0.0060 | -0.0004 | 0.9333 | 62.96 |
| 4 | 0.0168 | 0.0039 | -0.0003 | 0.8766 | 31.96 |
| 8 | 0.0125 | 0.0059 | -0.0006 | 0.8928 | 24.98 |
| 12 | 0.0190 | 0.0045 | -0.0004 | 0.9328 | 55.49 |
| 18 | 0.0112 | 0.0051 | -0.0005 | 0.9146 | 26.77 |
| 24 | 0.0106 | 0.0056 | -0.0006 | 0.9765 | 101.93 |

TABLE 7. Percentage relative efficiencies of plots of different sizes in blocks of different sizes and without blocking

| Block <br> sizes | Number of basic units ( x ) in a plot |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 6 | 8 | 9 | 12 |
| Without |  |  |  |  |  |  |  |  |
| blocking | 100 | 57.03 | 41.41 | 32.81 | 23.44 | 18.75 | 17.19 | 13.28 |
| 2 | 100 | 55.50 | 39.71 | 31.10 | 22.01 | 17.22 | 15.79 | 12.46 |
| 4 | 100 | 56.15 | 40.11 | 31.55 | 22.46 | 17.65 | 16.04 | 12.30 |
| 8 | 100 | 54.40 | 38.46 | 29.67 | 20.88 | 16.48 | 14.84 | 11.54 |
| 12 | 100 | 54.76 | 38.10 | 29.76 | 20.83 | 16.07 | 14.29 | 11.31 |
| 18 | 100 | 55.00 | 38.75 | 30.00 | 21.25 | 16.25 | 15.00 | 11.88 |

TABLE 8. Optimum plot sizes estimated from the Smith's equation by modified maximum curvature method

| Number of plots in a block | Optimum plot <br> sizes (in units) | Area of the plot ( $\mathrm{m}^{2}$ ) |
| :---: | :---: | :---: |
| Without blocking | 8.100 | 3.65 |
| 2 | 4.540 | 1.84 |
| 4 | 5.198 | 2.34 |
| 8 | 4.443 | 1.98 |
| 12 | 4.618 | 2.07 |
| 24 | 5.504 | 2.48 |

TABLE 9. Effect of size and shape of plots on coefficients of variation for blocks of different sizes and without blocking

| Plot dimensions |  | Size of blocks |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Size | Shape L $x$ B | Without blocking | 2 | 4. | 8 | 12 | 18 |
| 2 | $1 \times 2$ | 72.63 | 54.84 | 56.42 | 57.95 | 61.26 | 64.15 |
|  | *2 $\times 1$ | 64.92 | 37.21 | 41.02 | 46.35 | 49.84 | 52.13 |
| 4 | $1 \times 4$ | 61.01 | 42.26 | 42.90 | 44.57 | 49.57 | 55.91 |
|  | $2 \times 2$ | 61.91 | 41.76 | 44.57 | 45.96 | 50.78 | 52.30 |
|  | * $4 \times 1$ | 56.38 | 20.80 | 31.81 | 37.70 | 41.37 | 47.97 |
| 6 | $1 \times 6$ | 55.62 | 36.59 | 41.94 | 45.01 | 48.14 | 50.51 |
|  | $2 \times 3$ | 55.11 | 31.80 | 36.62 | 44.28 | 45.11 | 45.21 |
|  | $3 \times 2$ | 56.83 | 40.25 | 41.86 | 42.14 | 47.21 | 51.91 |
|  | *6 $\times 1$ | 48.69 | 24.54 | 26.05 | 29.26 | 31.47 | 36.66 |
| 8 | *1 $\times 8$ | 52.45 | 32.68 | 31.96 | - | 42.49 | 50.07 |
|  | $2 \times 4$ | 54.02 | 33.65 | 35.12 | - | 44.48 | 48.26 |
|  | $4 \times 2$ | 54.97 | 32.95 | 37.01 | - | 45.72 | 52.26 |
| 12 | $1 \times 12$ | 48.47 | 38.88 | 36.82 | 42.18 | 43.42 | 47.15 |
|  | $2 \times 6$ | 50.24 | 30.97 | 39.82 | 44.46 | 45.54 | 44.99 |
|  | $3 \times 4$ | 50.94 | 35.30 | 35.65 | 36.47 | 42.63 | 49.79 |
|  | $4 \times 3$ | 50.65 | 27.97 | 33.80 | 44.97 | 43.27 | 42.62 |
|  | $6 \times 2$ | 47.63 | 27.97 | 29.01 | 31.01 | 38.29 | 41.91 |
|  | *12.x 1 | 43.55 | 11.13 | 18.16 | 23.31 | 26.81 | 33.45 |

$\mathrm{L}=$ length (number of unit plot)
$\mathrm{B}=$ breadth (number of unit plot)

* The shape which has minimum coefficient of variation for particular plot size.

TABLE 10. Fercentage relative efficiencies of blocks of different sizes compared with without blocking for plots of different slzes

| Plot sizes |  | Size of blocks |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Units | $\begin{gathered} \text { Area } \\ \mathrm{m}^{2} \text { ) } \end{gathered}$ | Without <br> blockin | 2 | 3 | 4 | 6 | 8 | 9 | . 12 | 18 |
| 1 | (0.45) | 100.00 | 244.74 | 173.25 | 188.17 | 163.39 | 169.69 | 158.29 | 150.97 | 141.41 |
| 2 | (0.90) | 100.00 | 216.07 | 173.67 | 183.68 | 164.24 | 172.33 | 152.71 | 144.33 | 138.89 |
| 3 | $(1.35)$ | 100.00 | 246.62 | 218.55 | 207.82 | 176.65 | 175.04 | 172.81 | 153.22 | 144.88 |
| 4 | (1.80) | . 100.00 | 238.78 | 191.90 | 206.45 | 167.06 | 194.43 | 152.83 | 159.68 | 132.46 |
| 6 | (2.70) | 100.00 | 256.61 | 205.68 | 201.27 | 176.52 | 175.81 | 170.66 | 132.98 | 135.89 |
| 8 | (3.60) | 100.00 | 262.40 | 184.40 | 240.00 | 190.43 | - | 133.90 | 137.35 | 118.96 |
| 9 | (4.05) | 100.00 | 207.76 | 210.12 | 177.94 | 163.71 | 150.17 | - | 125.96 | - |
| 12 | (5.40) | 100.00 | 255.38 | 194.64 | 204.02 | 161.08 | 162.46 | 143.08 | 130.93 | 118.91 |
| 18 | (8.10) | 100.00 | 216.08 | 189.97 | 162.29 | 152.84 | 127.38 | - | 110.23 | - |

TABLE 11. Percentage relative efficiencies of blocks of different shapes compared with without blocking for plots of different sizes

| Block dimensions |  | Size of plots |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Size | Shape <br> L $\times$ B | 2 | 4 | 6 | 8 | 12 |
| Without |  |  |  |  |  |  |
| blocking | - | 100.00 | 100.00 | 100.00 | 100.00 | 100.00 |
| 2 | $1 \times 2$ | 243.50 | 254.13 | 286.39 | 234.63 | 231.86 |
|  | $2 \times 1$ | 194.20 | 218.95 | 232.44 | 297.61 | 250.30 |
| 4 | $1 \times 4$ | 185.15 | 225.95 | 188.94 | 211.65 | 160.61 |
|  | $2 \times 2$ | 198.67 | 187.00 | 223.38 | 235.11 | 187.29 |
|  | $4 \times 1$ | 157.42 | 214.04 | 188.58 | 283.91 | 222.78 |
| 8 | $1 \times 8$ | 206.94 | 209.95 | 180.67 | - | 184.48 |
|  | $2 \times 4$ | 172.33 | 169.39 | 175.81 | - | 161.08 |
|  | $4 \times 2$ | 147.64 | - | 171.21 | - | 152.24 |
| 12 | $1 \times 12$ | 137.07 | 148.23 | 166.08 | 118.98 | 128.14 |
|  | $2 \times 6$ | 159.66 | 133.33 | 154.96 | 124.70 | 119.73 |
|  | $3 \times 4$ | 155.43 | 171.99 | 136.75 | 166.24 | 134.19 |
|  | - $4 \times 3$ | 139.20 | 161.39 | 149.27 | 147.49 | 133.51 |
|  | $6 \times 2$ | 150.59 | 151.95 | 168.37 | 177.63 | 143.64 |
|  | $12 \times 1$ | 125.90 | 156.00 | 174.67 | 198.75 | 196.30 |

$\mathrm{L}=$ length (number of plots)
$B=$ breadth (number of plots)

| TABLE 12. | Relative efficiencies of blocks of different sizes without considering the size of the plot using Smith's Variance Law |  |
| :---: | :---: | :---: |
|  | Relative efficiencies |  |
| *P Vs Q | Using average Smith's index of soil heterogeneity, b | using Bmith's Index of soil heterogenelty, b; of without blocking |
| 4 Vs 2 | 0.8280 | 0.8505 |
| 6 Vs 2 | 0.7532 | 0.7849 |
| 8 Vs 2 | 0.7087 | 0.7458 |
| 12 Vs 2 | 0.6558 | 0.6993 |
| 6 Vs 4 | 0.9096 | 0.9228 |
| 8 Vs 4 | 0.8558 | 0.8768 |
| 12.Vs 4 | 0.7920 | 0.8222 |
| 8 Vs 6 | 0.9409 | 0.9502 |
| 12 Vs 6 | 0.8707 | 0.8910 |
| 12 Vs 8 | 0.9255 | 0.9377 |

* Block of "size $p$ compared with block of size 0

TABLE 13 Minimum number of replications ( $r$ ) and total experimental area (a) required for estimating treatment means with 5 percent standard error

| Block sizes |  | Plot sizes (m) |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 0.45 | 0.9 | 1.8 | 2.7 | 3.6 | 5.4 |
| Without blocking | $r$ | 244. | 186 | 142 | 122 | 109. | 93 |
|  | a | 109.8 | 167.4 | 255.6 | 329.4 | 392.4 | 502.2 |
| 2 | r | 91 | 74 | 60 | 53 | 48 | 42 |
|  | a | 40.95 | 66.6 | 108 | 143.1 | 172.8 | 226.8 |
| 4 | r | 11.4 | 91 | 73 | 64 | 58 | 51 |
|  | a | 51.3 | 81.9 | 131.4 | 172.8 | 208.8 | 275.4 |
| 8 | I | 121 | 101 | 85 | 77 | 71. | 64 |
|  | a | 54.45 | 90.9 | 153 | 207.9 | 255.6 | 345.6 |
| 12 | r | 141 | 119 | 101 | 91 | 85 | 77 |
|  | a | 63.45 | 107.1 | 181.8 | 245.7 | 306 | 426.8 |
| 18 | $r$ | 157 | 130 | 108 | 97 | 90 | 81 |
|  | a | 70.65 | 117 | 194.4 | 261.9 | 324 | 437.4 |

TABLE 14a. Percentage relative efficiencies of Latin Square Design over Randomized Block
Design

TABLE 14b. Percentage relative efficiencies of Lating Square Design over Completely Randomized Design

| Plot dimensions |  | $\begin{gathered} \text { Order of } \\ \text { ISD } \end{gathered}$ |  | Percentage relative efficiencies |
| :---: | :---: | :---: | :---: | :---: |
| Size | Shape <br> L $\boldsymbol{x}$ B |  |  |  |
| 6 | $1 \times 6$ | $12 \times 12$ |  | 204.78 |
|  | $2 \times 3$ | $6 \times 6$ | A | $\begin{aligned} & 193.46 \\ & 271.78 \end{aligned}$ |
|  | $1 \times 6$ | $4 \times 4$ | A | $\begin{aligned} & 386.80 \\ & 354.38 \end{aligned}$ |
|  | $3 \times 2$ | $4 \times 4$ | A | $\begin{aligned} & 140.60 \\ & 354.38 \end{aligned}$ |

[^0]TABLE 15. Percentage relative efficiencies of Randomized Block Design over Completely Randomized Design

| Plot dimensions |  | Percentage relative efficiencies |  |
| :---: | :---: | :---: | :---: |
| Size | Shape $\mathrm{Lx} \text { x }$ | Row as blocking | Column as blocking |
| 6 | $1 \times 6$ | 101.94 | 184.04 |
|  | $3 \times 2$ | 100.41 | 172.37 |
|  | $2 \times 3$ | 101.66 | 184.83 |

## Fra:IFERTIITY CONTOUR MAP

schas



A - mphastratul block dr kav
日. COLEGE of hogticirture
c - LADIES HOSTEL
D- expenmental hot
E F fost-ball onduris
postmon of the fislo

Fig:2: effect of plot size on yariabuity


## DISCUSSION

The determination of size and shape of experimental units or plots and their arrangements in groups or blocks of suitable size is of great importance in field experimentation. As the magnitude of experimental error depends largily on the dimensions of the experimental unit a clear insight on the proper size and shape of experimental unit is of immense use in increasing the efficiency of field experimentation. Optimum size and shape of the experimental units have been determined statistically for most of the field crops: But no such studies are known to have been reported on turmeric, an important commercial crop of India; At present field trials on crop improvements and agronomic practises on turmeric are being conducted using plots of widely different sizes and shapes depending on the availability of resources and practical convenience of the research worker. Therefore a uniformity trial was laid out at the experimental farm at the college of Horticulture, Vellanikicara with the main purpose of determining the suitable size and shape of experimental plots in conducting field trials on turmeric and the results of the trial are discussed here under various sections.
5.1. Soil Heterogeneity and Fertility Contour Map

The fertility contour map of the experimental field revealed that there were no specific trends of fertility variation in the field. On the whole the field can be considered to be heterogeneous. But for small plots homogeneity can be maintained and therefore smaller, plots arranged in blocks of relatively small size are expected to be more efficient than large plots arranged im blocks of relatively large size. As the variation in soil fertility of the field appeared to be patchy it would be better to use multinway classified designs, such as latin square, Youden square etc. From a uniformity trial on brinjal conducted on an adjacent field at Vellanikkara, Hariharan (1981), obtained similar results on the distribution of soil heterogeneity. Results also indicated the necessity of proper orientation of plots and blocks. If randomised block design is to be used orientation of blocks should be of prime consideration for the reduction of experimental error.

### 5.2. Estimation of Optimum Plot Size

Coefficient of variation (cv) for plots of different sizes and shapes were found to decrease with an increase in plot size but the decrease was not proportional. This finding appears to be an aspect of the general law relating to size of the plot and variabllity and has been inaccordance with all
the previous findings on the same line. It was also seen that reduction in cv in the North-South direction (column) was more rapid than that in the East-West direction (row). This may be due to silight slape in the field in the NorthSouth direction.
5.2.1. Method of Maximum Curvature (Free-hand Curve)

The optimum plot size estimated from the free-hand curves by the method of maximum curvature is about 6 units for blocks of various sizes and 8 undts for ungrouped data. These values were almost in agreement with the present popular plot size for turmeric viz. $3.6 \mathrm{~m}^{2}$. The results obtained through free-hand curve method and mathematical method were not very much different. But mathematical method indicated the possibility of further reduction in plot size than that obtained through free-hand method.
5.2.2. Smith's Law and Modified Maximum Curvature Method

Smith's equation in the modified form gave a satisfactory fit to the data in both the cases where the plots were grouped. into blocks of suitable sizes and there was no blocking. The estimated values of 'b', the Smith's Index of Soll Heterogeneity, were nearer to zero. The result indicated that there was strong correlation between contiguous units (Smith, 1938). Hande et al (1982) obtained similar values
of 'b' in their studies on oat. The study revealed the need for grouping plots into blocks of appropriate sizes for effective error control. As most of the variation is positional than genetic, direct methods of controlling error are of great importance than that of the indirect methods especially through covariance techniques. Results obtained from studies on various other annual crops also are in agreement with these findings.

### 5.2.3. Alternate Models to Sinith's Law

The three non-linear models other than that due to Smith's also gave promising results: Among them the equation $Y=a+b / \sqrt{x}+c / x$ was found to be the best choice. In most of the cases this equation was an improvement to the familiar Smith's equation in the modified form. But unlike Smith's function the parameters of the function cannot be attributed to any physical meaning. But these models can be conveniently utilised for estimating optimum plot size by various methods. Lessman and Atkins (1963a) found the function $\log Y=\frac{a}{(a+\log x)^{b}}$. was an improvement over Smith's function in describing the proper relationship between plot size and variability. In this study the three functions decribed here were found to be at least as efficient as the Smith's function.

### 5.2.4. Optimum plot Size from Smith's Function by considering Cost of Experimentation

As mentioned earlier, size of a plot is also governed by cost of experimentation. Looking at this problem from the angle of economy, the plot size which gives maximum information per unit cost would be considered to be optimum for a given experiment. Hence optimum plot size was worked out.by assuming various arbitrary values for the cost components of an assigned law. Saxana et al (1972) on o免t, …: Prabhakaran and Thomas (1974) on tapioca; Biswas et al (1982) on cabbage and Binns et al (1983) on tobacco have followed the same procedure. The results showed that plots of smaller size are more efficient than larger ones in case the cost ratio $K_{1} / K_{2}$ is less than inity. Thus for an experimenter with limited resources it would be always advantageous to select the smallest possible size of the plot where agricultural operations can be conveniently carried out for the conduct of the experiment. The loss In precision due to the use of such smaller plots will be negligible when compared to the overall saving of experimental material and other resources.
5.2.5. Modified Maximum Curvature Method

The optimum plot size was also determined from the Smith's equation mathematically by maximising the radius
of curvature of the Smith's curve. An expression for estimating the optimum was derived using differential calculus and it was further used for locating the optimal point. The result indicated that plot sizes in the range from $1.9 \mathrm{~m}^{2}$ to $3.7 \mathrm{~m}^{2}$ were optimal with blocks of various sizes and without blocking. As a single overal estimate, plots of size $2.7 \mathrm{~m}^{2}$ can be considered to be optimal. Thus, If sufficient resources are available the experimenter may use plot of size $2.7 \mathrm{~m}^{2}$ or $3 \mathrm{~m}^{2}$ for conducting field trials on tumeric. With the use of local control size of the plot can be further reduced to $2 \mathrm{~m}^{2}$ or less. optimum plot size detemined by the above technique are expected to be stable and produce consistent resultsin the long run. The recommended plot size for turmeric as mentioned above is closer to the existing popular plot size for turmeric viz. $3.6 \mathrm{~m}^{2}$. Thus there was no need for increasing the size of the plot beyond $3.0 \mathrm{~m}^{2}$ but it can be further reduced to $2 \mathrm{~m}^{2}$ oreven less without any appreciable loss in precision. The estimate of plot size obtained here is in agreement with that of radish suggested by Raghavarao (1983) who worked on the same lines.
5.2.6. Concept of Percentage Relative Efficiency

The percentage relative efficiency decreased as plot size increased for blocks of different sizes and without
blocking. As a rule small plots were found to be more efficient than large ones and the most efficient plot size was that with a single basic unit ( $0.45 \mathrm{~m}^{2}$ ). The result also was in close agreement with that of Menon and Tyagi (1971) on Mandarin orange; Bkurghava and Sardana (1973) on apple; and Prabhakaran et al (1978) on banana. Thus, if the cost of the experimentation is proportional to the population of plants or area of the experimental plots it would be beneficial to use the smallest possible plot size. But for crops like turmeric such assumption is far from true. Further, with very small plots agronomic operations cannot be carried out with added convenience.

### 5.3. Shape and Orientation of the Plots

In general plot shape did not seem to exert any consistent effect on cv . However, for a given plot size long and narrow plots gave lower cv than approximately square plots. This result was supported by Sreenath (1973) on sorghum; Prabhakaran and Thomas (1974) on tapioca and Hariharan (1981) on brinjal. The findings of Cochran (1940) that long and narrow plots have better control of error than a square plot are also on the same side.

Orientation of plots in a block is very 1mportant in deciding the efficiency of field experimentation. Proper orientation of plots was found to result in internal
homogeneity of the blocks and subsequent reduction in experimental error. In general orientation $b \times a$. where ' $b$ ' is the number of units in the column wise (North-South direction) and 'a' is the number of unit in the row wise (East-West direction) was found to be better than the orientation $a \times b(b) a)$.
5.4.

Size, Shape and Orientation of Blocks

Block efficiency was found to decrease with an increase in the number of plots per block. Similar results on other crops have been reported by Gopini et al (1970) on groundnut; Saxana et al (1972) on oat; Sreenath (1973) on sorghum; Bist et al (1975) on potato; Hariharan (1981) on brinjal and Nair (1981) on cashew. Two plot blocks were found to be the most efficient ones. The result called for the use of incomplete block designs in laying out field trials. Size and shape of plots in blocks did not exert any appreciable effect on block efficiency. This may be due to the fact that homogeneity of the plots can be achieved in smaller blocks even by using relatively large plots. The result is also in agreement with several earlier findings.

Orientation of block was also important in controlling error variation. From the study it was seen that orientation $b \mathrm{x}$ 'a, where ' b ' is the number of units in the column wise
(North-South direction) and ' $a$ ' is the number of unit in the row wise (East-West direction) was found to be better than the orientation $a \times b(a>b)$. That is the orientation of blocks perpendicular to the direction of the gradient and that of plots parallel to the gradient greadvantageous.

### 5.5. Number of Replications

For a fixed area of land, large number of replication with smallest possible plot size was found tó give lower standard error than smaller number of replications with relatively large plots. Thus it was more beneficial to use smaller plots with adequate number of replicatesthan large plots with fewer number of replications. The findings of Iyer and Agarwal (1970) on sugarcane; Bist et al (1975) on potato: Prabhakaran et $\underline{\text { al (1978) on banana and Suman and }}$ Wahi (1982) on cabbage are in confirmity with this result.

## 5.6. Efficiency of Experimental Design

In general Latin Square Design (LSD) was more efficient than Randomised Block Design (RBD) and Completely Randomised Design (CRD) . Similar resuits have been reported by Malhotra et al (1979.) on potato.

Randomised Block Designs with columns as blocks was also found to be equally efficient with Latin Square Designs. The result indicated the need for proper orientation of plots
and blocks in Randomised Block Designs to make the design as efficient as Latin Square Designs. If nothing is known about the direction of the fertility gradient it would be better to use Latin Square Design. The results are in agreement with the findings of Jayaraman (1979) on sunflower.

SUMMARY

A uniformity trial on turmeric was conducted at the experimental field of the College of Horticulture, Vellanikkara during the period from June 1983 to January 1984. At the time of harvest, the yield data from 864 plots each of size $0.6 \mathrm{~m} \times 0.75 \mathrm{~m}$ were recorded separately, discarding the external border row. The salient results of the statistical analysis of the uniformity trial data are given below.
6.1 The fertility contour map of the field showed that there was appreciable variation in soil fertility but this variation did not follow any systematic pattern. As a matter of fact, small areas were relatively more homogeneous with regarde: to soil fertility than large areas.
6.2 An increase in the plot size in either direction decreased the coefficient of variation, but the decrease was not proportional. Further, the reduction in cv in the North-South (column) direction was more rapid than that in the East-West (row) direction.
6.3 The empirical law suggested by Smith (1938) gave a satisfactory fit to the data for blocks of different sizes and without blocking. The empirical models suggested by Prabhakaran were found to be more efficient than Smith's Eunction.
6.4. The optimum plot sizes estimated through Smith's index of soil heterogeneity method, maximum curvature method and modified maximum curvature method, were not much different. For a general recommendation a plot size $2.7 \mathrm{~m}^{2}$ ( $3.6 \mathrm{~m} \times 0.75 \mathrm{~m}$ ) or approximately $3 \mathrm{~m}^{2}$ was found advisable for conducting field trials on turmeric. but for with block designs the plot size can be reduced even to $2 \mathrm{~m}^{2}$ without. much loss in overal precision of treatment comparison.
6.5. The shape of the plot did not exert any consistent effect on coefficient of variation. However, long and narrow plots gave lower ev than approximately square plots in most situations.
6.6. Efficiency of blocking may be considered to be a function of the block size. Two plot blocks were the most efficient in controlling error.
6.7. The shape of the blocks had no consistent effect on the variability whereas proper arrangement of plots and blocks resulted in a considerable reduction of experimental error.
6.8. An increase in plot size was followed by a decrease in the expected number of replications per treatment but the decrease was not proportional. Increasing the number of replications rather than plot size was found to be more advantageous for the enhancement of precision.
6.9. In general Latin Square Design was found to be more efficient than Randomised Block Design and Completely Randomised Design $\dagger$ But with proper arrangement of blocks and plots within the block, the efficiency of Rendomised Block Design can be considerably increased.

REFERENCES

Agarwal, K.N.; Bavappa, K.V.A. and Khosla, R.K. (1968) Study of size of plots and blocks and number of pre-experimental periods of arecanut. Indian. ${ }^{\text {. }}$ Agric. Sci. 38: 444-460.

Baker, G.A.: Huberty, M.R. and Veihmeyer, F.J. (1952). A uniformity trial on unirrigated barley of ten year's duration. Agron.J. 44: 267-270.

Bhargava, P.N. and Sardana, M.G. (1973). Size and shape of plots in Eleld trials with apple. Indian. Hort. 32: 50-57.

Binns, M. Ian Ogilive.: Neislle Arnold and Petras Lukosevicius. (1983). Size of plots and blocks in Agricultural experiments. An empirical study with cigar-filler tobacco. Indian.J. Agri. Sci. 53 (8) : 706-718.

Bist, B.S.; Malhotra, V.P. and Sreenath, P.R. (1975). Size and shape of plots and blocks in field experiments with potato crops in hills. Indian. Sc1. $\frac{45}{\underline{=}(1)}$ : 5-8.

Biswas, S.R.; Ramachandran, P.R. and Gupta, A. (1982). Size and shape of experimental plots in cabbage experiments. Indian. U. Agric. Sci. $5 \underline{\underline{5} 2(\underline{\underline{2}}): 1939-1941 . ~}$
*Bose, R.C. (1935). Some soll heterogenelty trials at Pusa and the size and shape of experimental plots. Indian. I. Agric: Sci. 5: 579-608.
*Chiristidis, B.G. (1931). The importance of the shape of plots in field experimentation. J. Agric. Sci. Camb. 21: 14-37.
*Cochran, W.G. (1937). Catalogue of undformity trial data. Suppl. J1. R. Statist. Sci. 色: 233-253.

Cochran,' W.G. (1940). A survey of experimental desion. Mimeo. USDA.

Federer, W.T. (1955). Experimental Designs. Theory and Application. Oxford \& IBH Publishing Co., New Delhi. Indian. Ed. pp. 58-59 and 86-163.

Fisher, R.A. (1926). Statistical methods for research workers. Uliver and Boyd, London. 1st Ed. pp. 1-198.

Fisher, R.A. (1951). The design of experiments. Hafner Publ. Co., Newyork, 6th Ed. pp. 1-244.

Freeman (1963). The combined effect of environmental and plant variation. 日iometric. 19: 273-277.
*Gomez, K.A. (1972). Techniques for Eield experiments with rice. The International Rice Research Institute, Philippines. pp. 1-18.

Gopini, D.D.; Kabaria, M.M. and Vaishnani, N.L. (1970). Size and shape of plots in field experiments on groundnut. Indian.J. Agric. Sci. $40(111): 1004-1010$.

Handa, D.P.; Sreenath, P.R.; Sastry, J.A.; Rajpal, S.K. and Shukla. (1982). Size and shape of plots and blocks for experiments with oat grown for fodder. Indian. I. Agric. Sci. $52(7): 466-471$.

Hariharan. (1981). Optimum plot size for field experiments on brinjal. Unpublished M.Sc. (Ag. Stat.) Thesis, Kerala Agricultural University, Trichur.
*Harris, J.A. (1915). On a criterion of substratum homogeneity in the field experiments. Am. Nat. 49: 430-454.
*Harris, J.A. (1920). Practical universality of field heterogeneity as a Eacter influencing plot yields. J.Agri. Res. 19: 279-314.

Hatheway. (1961). Convenient plot size. Agron. U. 5 . 3 : 279-280.

Hatheway and williams (1958). Efficient estimation of relationship between plot size and the variability of crop yields. Biometrics. 14: 207-222.
*Hayes. (1925). Control soil heterogeneity and use of the probable error concept in plant breeding studies. Minn. AES. Tech. BuIl. 30: 1-21:
*Hodnett, G.E. (1953). A undformity trial on groundnuts. J. Agric. Sci. Comb. 43: 323-328.
*Hutchinson, J.B. and Panse, V.G. (1935a). Studies in the Techniques of fleld experiments, size, shape and arrangement of plots in cotton trials. Indian. I. Agric. ScI. 5: 1-14.

Iyer, S.S. and Agarwal. K.C. (1970). Optimum size and shape of plots for sugarcane. Indian. 124-139.

Jayaraman. (1979). Optimum size and shape of plots and blocks and relative efficiency of design for Eleld experiments in sunflower (Heilanthus Annus.L) Unpublished M.Sc. (Ag.Stat.) Thesis, University of Agriculture Science, Bangalore.
*Kadam, B.S. and Patel, S.M. (1937). Studies in Field plot techniques with bajari (P. typhoideum.) Emp. I. Exp. Agric. 5 홍 219-230.
*Kalamkar, R.J. (1932a). Experimental error and the field plot technique with potatoes J. Agric: Sci. Camb. 22 : 373-385.

Katyal. V. and Sasmal, B.C. (1982). Influence of size and shape of plot and blocks on experimental error in field experiments conducted with jute (JRC-212) Indian. Agric. $26(1): 43-49$.

Koch, E.J. and Rigney, H.J. (1951). A method of estimating plot size from experimental data. Agron. J. 43: 17-21.

Kripashankar, M.S.: Lal, M.S. and Goswami, G. (1972). Size and shape of plots and blocks in yield trials of soyabean (Glycine Max.L.Merr.). Indian. Sc1. $42(10): 901-904$.
*Lessman, K.J. and Atkins, R.E. (1963a). Optimum plot size and relative efficiency of latice designs for grain sorghum yield tests. Crop. Sci. 3: 477-481.

Malhotra, V.P.: Bist, B.S. and Sreenath, P.R. (1979). Efficiency of Latin Square Design for experiments with potato crops in hills. Indian. I. Agric. Sci. 49 (8) : 655-658。

Marani. A. (1963). Estimation of optimum plot size using Smith's procedure. Agron. I. 55: 503
*Marcer, W.B. and Hall, A.D. (1911). The experimental error of field trials. J. Agric. Sci. Camb. $\frac{4}{\underline{3}}$ : 107m132.

Menon, T.C.M. and Tyagi. B.N. (1971). Optimum size and shapeof plots in experiments with mandarin orange. Indian-J. Agric. Sci. 49: 857-861.

Nair, R.B. (1981) Determination of size and shape of plots for trials on cashew. Unpublished M.Sc.(Ag.Stat.) Thesis, Kerala Agriculture University, Trichur.
*Pan, C.I. (1930). Unformity trials with rice. J. Amer. Soe. Agron. 27: 279-285.
*Panse, V.G. (1941). Studies in the techniques of field experiments. Size and shape of blocks and arrangement of plots in cotton trials. Indian.I. Agric. Sci. 11: 850-865.

Prabhakaran, P.V. (1983). An alternate approach for the estimates of plot size. Fnpublished. Personal Communteactor. Jounal of Kevala statistical Asfuaction 5 $\$ 24-34$

Prabhakaran, P.V.; Balakrishnan, S. and Mary George (1978). Optimum plot size for field trials with banana. Agric. Res. J. Kerala. 16: 33-38.

Prabhakaran, P.V. and Thomas, E.J. (1974). Optimum plot size for field experiments with tapioca. Agric. Res. U. Kerala. 12: 19-23.

Raghavarao. D. (1983) . Statistical technigues in agricultural and biological research. Oxford \& IBH Publishing Co.. 1st Ed. pp. 198-199.

Rambabu; Aarwal, M.C.; Somaraj, P. and Cninnamant, S. (1980). Size and shape of plots and blocks for field trials on natural grass lands in Nilgiris hills. Indian. J. Agric. Sci. 50: 598-602.

Saxana, P.N.: Kavitkar, A.G. and Monga, M.K. (1972). Optimum plot size for oat grown for fodder. Indian. J. Agric. Sci. 42 (1) : 63-69.
*Sen. A.R. (1963a). Use of pre-treatment data in designing experiments on tea. Emp. ${ }^{\text {. Exp }}$ Agric. 31: 41-49.
*Smith, H.F. (1938). An empirical law describing heterogeneity in the yields of agricultural crops. U. Agric. Sci. Comb. 28: 1-23.

Sreenath, P.R. (1973). Size and shape of plots and blocks in field trials with "M.P.Chari" sorghum. Indian. Agric. Sci. $43(\underline{2}$ ) : 110-112.

Suman, C.I. and Wahi, S.D. (1980). Size and shape of plot for entomological experiments on cabbage leaf webber (Crocidolomia Binotalis). Indian. U. Agric. Res. 16 (4) : 261-264.
*Taylor, F.W. (1907-1909). The size of the experimental plot for field crops. Proc. An. Soc. Agron. 1: 56-58.
*Originals are not referred.

APPENDICES

## APPENDIX I - THE YIELD DATA

Weight of tumeric in grams.

| C | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| R |  |  |  |  |  |  |  |  |  |  |  |  |
| 1 | 155 | 195 | 110 | 305 | 450 | 225 | 437 | 283 | 300 | 186 | 139 | 133 |
| 2 | 25 | 40 | 14 | 67 | 157 | 260 | 283 | 707 | 500 | 134 | 237 | 389 |
| 3 | 323 | 615 | 98 | 20 | 125 | 510 | 140 | 295 | 129 | 232 | 609 | 370 |
| 4 | 25 | 75 | 0 | 0 | 205 | 95 | 167 | 150 | 276 | 153 | 385 | 831 |
| 5 | 502 | 807 | 202 | 127 | 95 | 160 | 640 | 490 | 310 | 490 | 278 | 987 |
| 6 | 105 | 505 | 207 | 205 | 143 | 330 | 286 | 235 | 115 | 231 | 251 | 83 |
| 7 | 515 | 507 | 78 | 140 | 0 | 0 | 262 | 269 | 219 | 337 | 225 | 162 |
| 8 | 0 | 26 | 0 | 0 | 140 | 20 | 96 | 252 | 323 | 323 | 600 | 692 |
| 9 | 30 | 30 | 0 | 0 | 310 | 540 | 771 | 340 | 282 | 280 | 384 | 329 |
| 10 | 100 | 78 | 100 | 30 | 0 | 40 | 214 | 291 | 292 | 825 | 547 | 245 |
| 11 | 76 | 35 | 113 | 280 | 0 | 0 | 219 | 447 | 258 | 162 | 278 | 387 |
| 12 | 0 | 50 | 0 | 45 | 0 | 0 | 351 | 607 | 385 | 605 | 313 | 384 |
| c | 13 | 14 | 15 | 16 | 17 | 18 | 19 | 20 | 21 | 22 | 23 | 24 |


| 1 | 784 | 618 | 415 | 438 | 234 | 304 | 973 | 495 | 1315 | 947 | 865 | 0 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 778 | 568 | 553 | 569 | 755 | 978 | 1095 | 930 | 736 | 797 | 370 | 855 |
| 3 | 387 | 630 | 850 | 503 | 902 | 1270 | 502 | 809 | 777 | 737 | 743 | 486 |
| 4 | 272 | 311 | 227 | 381 | 1250 | 871 | 267 | 122 | 402 | 321 | 435 | 222 |
| 5 | 248 | 555 | 238 | 282 | 660 | 826 | 550 | 439 | 930 | 507 | 457 | 460 |
| 6 | 388 | 288 | 432 | 650 | 468 | 876 | 510 | 563 | 318 | 210 | 675 | 450 |
| 7 | 272 | 302 | 232 | 130 | 272 | 434 | 121 | 97 | 80 | 85 | 220 | 143 |
| 8 | 382 | 418 | 377 | 304 | 401 | 429 | 378 | 176 | 490 | 143 | 560 | 617 |
| 9 | 233 | 318. | 270 | 335 | 509 | 373 | 27 | 152 | 160 | 215 | 540 | 603 |
| 10 | 854 | 763 | 396 | '831 | 362 | 351 | 960 | 595 | 152 | 407 | 87 | 40 |
| 11 | 367 | 447 | 802 | 660 | 1.054 | 1340 | 203 | 66 | 20 | 176 | 8 |  |
| 12 | 352 | 723 | 500 | 940 | 807 | 775 | 644 | 802 | 0 | , | 118 | 129 |


| $\mathrm{R}^{\text {C }}$ | 25 | 26 | 27 | 28 | 29 | 30 | 31 | 32 | 33 | 34 | 35 | 36 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 429 | 289 | 432 | 275 | 273 | 292 | 837 | 856 | 689 | 727 | 396 | 496 |
| 2 | 80 | 122 | 300 | 305 | 107 | 80 | 619 | 1163 | 259 | 453 | 424 | 330 |
| 3 | 280 | 333 | 200 | 300 | 254 | 310 | 741 | 500 | 594 | 375 | 318 | 318 |
| 4 | 344 | 510 | 676 | 409 | 373 | 361 | 762 | 1040 | 550 | 622 | 1100 | 355 |
| 5 | 379 | 468 | 989 | 965 | 498 | 834 | 178 | 238 | 625 | 119 | 588 | 670 |
| 6 | 310 | 400 | 877 | 600 | 528 | 607 | 283 | 528 | 515 | 829 | 560 | 237 |
| 7 | 0 | 0 | 314 | 103 | 738 | 824 | 226 | 203 | 463 | 265 | 524 | 748 |
| 8 | 111 | 137 | 495 | 1068 | 290 | 400 | 223 | 350 | 1028 | 703 | 417 | 559 |
| 9 | 274 | 110 | 570 | 670 | 427 | 303 | 375 | 404 | 230 | 407 | 895 | 995 |
| 10 | 0 | 35 | 657 | 661 | 171 | 324 | 524 | 290 | 945 | 379 | 887 | 530 |
| 11 | 0 | 0 | 344 | 236 | 525 | 276 | 770 | 1143 | 469 | 498 | 1247 | 1204 |
| 12 | 0 | 0 | 417 | 346 | 250 | 849 | 809 | 1255 | 754 | 823 | 365 | 780 |

(ii)

Weight of turmeric in grams

| C | 37 | 38 | 39 | 40 | 41 | 42 | 43 | 44 | 45 | 46 | 47 | 48 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| R |  |  |  |  |  |  |  |  |  |  |  |  |
| 1 | 570 | 705 | 494 | 296 | 1207 | 1310 | 1133 | 1006 | 1158 | 500 | 1107 | 1017 |
| 2 | 885 | 628 | 400 | 249 | 1300 | 1375 | 1220 | 1520 | 123 | 123 | 772 | 1034 |
| 3 | 419 | 624 | 767 | 434 | 1423 | 1042 | 1295 | 891 | 825 | 1333 | 365 | 60 |
| 4 | 1262 | 967 | 1058 | 925 | 1620 | 997 | 972 | 890 | 1245 | 1475 | 557 | 1210 |
| 5 | 497 | 940 | 1175 | 1225 | 709 | 230 | 179 | 175 | 965 | 1417 | 814 | 630 |
| 6 | 232 | 564 | 209 | 108 | 344 | 580 | 192 | 277 | 705 | 396 | 1112 | 450 |
| 7 | 344 | 439 | 328 | 92 | 400 | 581 | 0 | 0 | 379 | 627 | 125 | 638 |
| 8 | 422. | 554 | 550 | 687 | 543 | 627 | 500 | 475 | 574 | 905 | 540 | 1110 |
| 9 | 1440 | 1032 | 964 | 669 | 1124 | 710 | 1608 | 1000 | 857 | 1518 | 8.14 | 340 |
| 10 | 325 | 181 | 1118 | 1032 | 1468 | 927 | 1130 | 1113 | 1113 | 1545 | 803 | 474 |
| 11 | 500 | 582 | 1017 | 1690 | 1102 | 1197 | 1264 | 1349 | 375 | 310 | 1600 | 1375 |
| 12 | 184 | 327 | 550 | 350 | 500 | 491 | 768 | 610 | 645 | 795 | 750 | 675 |


| C | 49 | 50 | 51 | 52 | 53 | 54 | 55 | 56 | 57 | 58 | 59 | 60 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| R |  |  |  |  |  |  |  |  |  |  |  |  |
| 1 | 133 | 370 | 500 | 735 | 553 | 307 | 860 | 608 | 413 | 545 | 1828 | 615 |
| 2 | 592 | 842 | 380 | 760 | 1212 | 997 | 430 | 467 | 570 | 542 | 828 | 240 |
| 3 | 430 | 290 | 562 | 398 | 626 | 1340 | 888 | 1215 | 370 | 279 | 1230 | 1112 |
| 4 | 705 | 940 | 530 | 358 | 532 | 1008 | 830 | 1119 | 335 | 267 | 613 | 534 |
| 5 | 1317 | 835 | 145 | 187 | 399 | 335 | 305 | 833 | 200 | 325 | 410 | 428 |
| 6 | 560 | 740 | 710 | 229 | 633 | 646 | 318 | 243 | 30 | 85 | 575 . | 555 |
| 7 | 600 | 620 | 223 | 50 | 260 | 63 | 680 | 1057 | 87 | 115 | 220 | 374 |
| 8 | 537 | 440 | 1175 | 1157 | 1015 | 1428 | 968 | 835 | 1450 | 868 | 1712 | 1580 |
| 9 | 802 | 1245 | 1153 | 1043 | 448 | 665 | 583 | 648 | 785 | 246 | 270 | 436 |
| 10 | 520 | 803 | 448 | 150 | 382 | 675 | 1185 | 981 | 137 | 590 | 130 | 80 |
| 11 | 1180 | 1015 | 225 | 142 | 92 | 170 | 676 | 417 | 348 | 166 | 50 | 0 |
| 12 | 247 | 259 | 770 | 392 | 491 | 623 | 597 | 175 | 905 | 1235 | 25 | 60 |


| $c$ | 61 | 62 | 63 | 64 | 65 | 66 | 67 | 68 | 69 | 70 | 71 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |

R

| 1 | 1298 | 453 | 668 | 900 | 280 | 231 | 0 | 0 | 496 | 473 | 0 | 0 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 1198 | 1133 | 152 | 145 | 485 | 525 | 0 | 0 | 390 | 370 | 0 | 0 |
| 3 | 1003 | 1629 | 750 | 361 | 262 | 373 | 880 | 822 | 270 | 200 | 230 | 568 |
| 4 | 585 | 617 | 893 | 618 | 573 | 668 | 514 | 1035 | 0 | 0 | 133 | 302 |
| 5 | 125 | 225 | 584 | 455 | 819 | 1012 | 350 | 255 | 0 | 0 | 0 | 150 |
| 6 | 783 | 515 | 235 | 100 | 669 | 568 | 895 | 635 | 0 | 0 | 0 | 0 |
| 7 | 35 | 170 | 177 | 342 | 865 | 1034 | 505 | 275 | 100 | 273 | 0 | 94 |
| 8 | 443 | 1237 | 616 | 988 | 635 | 309 | 620 | 624 | 440 | 542 | 0 | 120 |
| 9 | 0 | 100 | 638 | 782 | 321 | 333 | 205 | 392 | 479 | 230 | 0 | 0 |
| 10 | 0 | 203 | 283 | 521 | 1018 | 830 | 175 | 40 | 0 | 0 | 0 | 0 |
| 11 | 35. |  | 0 | 50 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 12 | 30 | 0 | 0 | 170 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

C - Columns.
R - Rows.

# OPTIMUM PLOT SIZE FOR FIELD EXPERIMENTS ON TURMERIC (CURCUMA LONGA. L.) 

## By

B. GOPAKUMARAN NAIR

ABSTRACT OF A THESIS<br>Submitted in partial fulfilment of the requirements for the degree of T) Raster of \$cience (Agricultural Statisfics)<br>Faculty of Agriculture<br>Kerala Agriculture University

Department of Statistics

Mannuthy, Trichur
1984

A uniformity trial on turmeric (Curcuma longa. L.) was conducted at the experimental field of College of Horticulture, Vellanikkara, during the period from June 1983 to January 1984 to assess the nature and magnitude of soil heterogenelty of the experimental field, and to determine the optimum size and shape of experimental plots and blocks in conducting field trials on turmeric by different methods. At the time of harvest, the yield data from 864 plots each of size $0.6 \mathrm{~m} \times 0.75 \mathrm{~m}$ were recorded separately, discarding the external border row.

The fertility contour map of the field showed that the experimental field was not homogeneous as far as the fertility variation was concerned, It was observed that an increase in the plot size in either direction decreased the cv but the reduction'in cv was not proportional.

The empirical law suggested by Smith (1938) gave a satisfactory fit to the data for blocks of different sizes and without blocking and those suggested by Prabhakaran(1983) were found to be better than the Smith's law.

As a general recommendation, the optimum plot size for conducting field trials on turmeric was found to be $2.7 \mathrm{~m}^{2}$, but with blocks of small sizes, the optimum plot size can be reduced to $2 \mathrm{~m}^{2}$ or even less. Shape of the plot did not exert
any consistent effect on cv. However, long and narrow plots gave lower cv than approximately square plots. Thus for $2.7 \mathrm{~m}^{2}$, the plot shape $3.6 \mathrm{~m} \times 0.75 \mathrm{~m}$ was found to be optimum,

As the size of the block Increased efficiency of blocking decreased. Two plot blocks were the most efficient ones. Shape of the block had no consistent effect on variam bility whereas proper arrangement of plots and blocks resulted in a considerable reduction of experimental error.

An increase in plot size was followed by a decrease in the expected number of replications but the decrease was not proportional. Increasing the number of replications was found to be more advantageous than that of increasing the plot size.

In general Latin Square Design (LSD) was found to be more efficient than Randomised Block Design (RED) and Completely Randomised Design (CRD). But by the proper orientation of plots and blocks in Randomised block Design was found to be as efficient as Latin Square Design,


[^0]:    A - Average
    B - Combined analysis by eliminating between sets of sum of squares of LS.

