

**SIMULTANEOUS OPTIMIZATION OF SEVERAL RESPONSE VARIABLES  
IN A GRANULATION PROCESS**

D. Vojnovic, M. Moneghini, F. Rubessa,

Department of Pharmaceutical Sciences

University of Trieste, 34127 Trieste, Italy

and

A. Zanchetta

Zanchetta s.r.l., 55010 Lunata, Lucca, Italy

**ABSTRACT**

Wet granulation of lactose and corn starch in a 10 litre high shear mixer was examined. The optimal combination of three independent variables (moisture level, impeller speed and granulation time) on four properties of the granules was investigated by simultaneous optimization method.

The optimum zone determined in 10 litre high shear mixer was analyzed for scaling-up study. The same product was manufactured at 50 L scale and optimum theoretical results were obtained for the four response variables and are comparable with the experimental results.

## INTRODUCTION

A common problem in product development involves the selection of a set of conditions, the  $X$ s, which will result in a product with a desirable combination of properties, the  $Y$ s, response variables.

This becomes a problem in the simultaneous optimization of the  $Y$ s, each of them depends upon a set of independent variables,  $X_1, X_2, \dots, X_k$ . One approach to this problem has been proposed by Hartmann and Beaumont(1), Nicholson and Pullen(2) who described optimization schemes based upon the linear programming model. Harrington(3) presented an optimization scheme utilizing what he termed the desirability function. Gatza and Mc.Millan(4) gave a slight modification of Harrington function.

We will employ the form of Derringer and Suich(5) and illustrate its use in a granulation process in a high-shear mixer. In addition, we will also plot this desirability function against two independent variables with the third one held at its optimum level. Furthermore this function will be used to propose a method for scaling-up study in a high-shear mixer.

Each of the  $k$  response variables is related to the  $p$  independent variables by the relation:

$$Y_{ij} = f_i(X_1, X_2, \dots, X_p) + e_{ij} \quad \begin{matrix} i=1, 2, \dots, k \\ j=1, 2, \dots, \eta_i \end{matrix}$$

where  $f_i$  denotes the functional relationship between  $Y_i$  and  $X_1, X_2, \dots, X_p$ . We note that this function may differ for each  $Y_i$  and that  $f_i$  represents this relationship except for an error term,  $e_{ij}$ . If we make the usual assumption that  $E(e_{ij})=0$  for each  $i$ , then we can relate the average or expected responses  $\eta_i$  to the  $p$  independent variables by

$$\eta_i = f_i(X_1, X_2, \dots, X_p) \quad i=1, 2, \dots, k.$$

In practice,  $f_i$  is unknown. The usual procedure is to approximate  $f_i$  by a polynomial function. We then estimate  $\eta_i$  by  $Y_i$ , obtained through regression techniques. The desirability function involves transformation of each estimated

response variable  $Y_i$  to a desirability value  $d_i$ , where  $0 < d_i < 1$ ; the value of  $d_i$  increases as the "desirability" of the corresponding response increases. The individual desirabilities are then combined using the geometric mean.

$$D = (d_1 \cdot d_2 \cdots d_k)^{1/k} \quad \text{eq.1}$$

This single value of  $D$  gives the overall assessment of the desirability of the combined response levels. Clearly the range of  $D$  will fall into the interval (0-1) and  $D$  will increase as the balance of the properties becomes more favorable.  $D$  also has the property that if any  $d_i = 0$  (that is, if one of the response variables is unacceptable) then  $D = 0$  (the overall product is unacceptable). It is for these reasons that the geometric mean, rather than some other function of the  $d_i$  such as the arithmetic mean, was used.

In transforming  $Y_i$  to  $d_i$  two cases arise: one-sided and two-sided desirability transformations.

#### One-sided transformations

For the one-sided case,  $d_i$  increases as  $Y_i$  increases and is employed when  $Y_i$  is to be maximized. Many transformations are possible and we shall consider the transformations given by:

$$d_i = \begin{cases} 0 & Y_i \leq Y_a \\ \left[ \frac{Y_i - Y_a}{Y_b - Y_a} \right] & Y_a < Y_i < Y_b \\ 1 & Y_i \geq Y_b \end{cases}$$

and graphed in Figure 1.

The value  $Y_a$  gives the minimum acceptable value of  $Y_i$ . The user specifies this value of  $Y_a$ , knowing that any lower value of  $Y_i$  would result in an overall unacceptable product, since  $Y_i < Y_a$  would make  $d_i = 0$ , and thus  $D = 0$ , which indicates an unacceptable product.

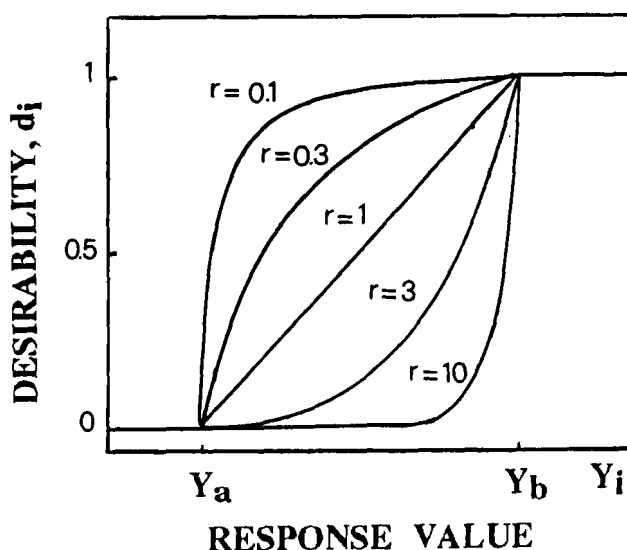


FIGURE 1

## One-sided desirability functions

The value  $Y_b$  gives the highest value of  $Y_i$ . Since we are considering a one-sided transformation, there is no highest value of  $Y_i$ . However, from a practical viewpoint, one can think of  $Y_b$  as the value for  $Y_i$  such that higher values of  $Y_i$  have little additional merit.

The value of  $r$  used in the transformation would again be specified by the user. Figure 1 indicates that a large value of  $r$  would be specified if it were very desirable for the value of  $Y_i$  to increase rapidly above  $Y_a$ . In other words, even though  $Y_a$  is an acceptable value the desirability of the product would be greatly increased by having  $Y_i$  considerably greater than  $Y_a$ .

As can be seen, the desirability  $d_i$  then increases slowly as  $Y_i$  increases. Therefore, to maximize  $d_i$  and thereby  $D_i$ ,  $Y_i$  must be greatly increased over  $Y_a$ . On the other hand, a small value of  $r$  would be specified if having values of  $Y_i$  considerably above  $Y_a$  were not of critical importance. A value of  $r=0.1$ , for example, would mean that any value of  $Y_i$  above  $Y_a$  was just about as desirable as any other value of  $Y_i$  above  $Y_a$ .

Two-sided transformations

The two-sided transformation arises when the response variable  $Y_i$  has both a minimum and a maximum constraint. We shall consider the transformations given by:

$$d_i = \begin{cases} \left[ \frac{Y_i - Y_a}{C_i - Y_a} \right]^s & Y_a \leq Y_i \leq C_i \\ \left[ \frac{Y_i - Y_b}{C_i - Y_b} \right]^t & C_i < Y_i \leq Y_b \\ 0 & Y_i < Y_a \quad \text{or} \quad Y_i > Y_b \end{cases}$$

In this situation  $Y_a$  is the minimum acceptable value of  $Y_i$  and  $Y_b$  is the maximum acceptable value. Values of  $Y_i$  outside these limits would make the entire product unacceptable. The value selected for  $c_i$  would be that value of  $Y_i$  which was most desirable and could be selected anywhere between  $Y_a$  and  $Y_b$ . The values of  $s$  and  $t$  in the two-side transformations play the same role as  $r$  does in the one-sided transformation.

In Figure 2 several different values of  $t$  and  $s$  are plotted. This figure also shows that large values for  $s$  and  $t$  would be selected if it were very desirable for the value of  $Y_i$  to be close  $c_i$ . In this case the desirability  $d_i$  would not get large until  $Y_i$  got close to  $c_i$ . On the other hand, if almost any value of  $Y_i$  above  $Y_a$  and below  $Y_b$  were acceptable, then small values of  $s$  and  $t$  would be chosen. Moderate values for  $s$  and  $t$  (near 1) would represent a compromise between the two extremes. One could also select a large value of  $s$  and a small value of  $t$  if it were desirable for  $Y_i$  to increase rapidly to  $c_i$  while almost any value of  $Y_i$  above  $c_i$  but below  $Y_b$  was also desirable.

The procedure outlined can be used to maximize some of the  $d_i$  (corresponding to certain  $Y_i$ ) while in putting constraints on the other  $Y_i$ , similar to a linear .For

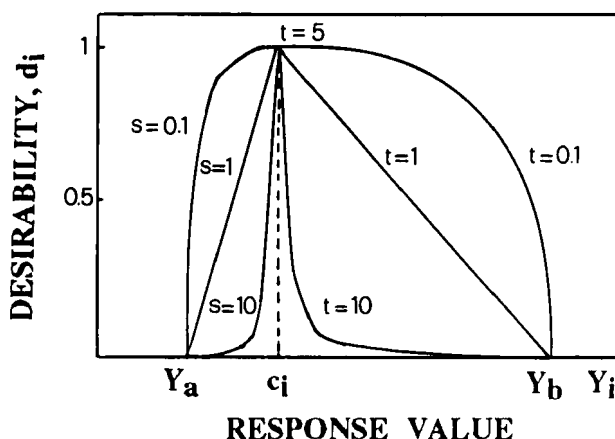


FIGURE 2

Two-sided desirability functions

those  $Y_i$  that are subject to constraints one uses extremely small values of the exponents ( $r$ ,  $s$ , and  $t$ ) and allows  $Y_a$  and  $Y_b$  to act as the boundary values.

### Methods of optimization

We have assumed that  $Y_i$  is a continuous function of the  $X_h$ . From eq.2 and 3 we see that the  $d_i$  are a continuous function of  $Y_i$  and from eq.1 that  $D$  is a continuous function of the  $d_i$ . Therefore, it follows that  $D$  is a continuous function of the  $X_h$ . As a result, existing univariate search techniques can be used to maximize  $D$  over the independent variable domain. In essence, the desirability function condenses a multivariate optimization problem into a univariate one. An added benefit of the method is the ability to plot  $D$  as a function of one or more independent variables.

## EXPERIMENTAL SECTION

**Materials-** Lactose(200 mesh, DMV, The Netherlands), corn starch(Gianni, Italy) and polyvinylpyrrolidone K 25(Gaf, Italy) were used as starting materials.

**Instrument-** The Zanchetta Roto J and Roto P granulators are similar in design to many vertical high speed mixer granulators having a large impeller.

TABLE 1  
Experimental design

N°	X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>
1	-1	-1	-1
2	+1	-1	-1
3	-1	+1	-1
4	+1	+1	-1
5	-1	-1	+1
6	+1	-1	+1
7	-1	+1	+1
8	+1	+1	+1
9	-1.68	0	0
10	+1.68	0	0
11	0	-1.68	0
12	0	+1.68	0
13	0	0	-1.68
14	0	0	+1.68
15	0	0	0
16	0	0	0
17	0	0	0
18	0	0	0
19	0	0	0
20	0	0	0

Granulate manufacture- A 2 kg and 10 kg batch of lactose (68.2%) and corn starch (31.8%) were granulated with an aqueous solution of povidone(2%).Impeller speed was set at 493 or 245 rpm respectively for Roto J and Roto P.The granulation time was 8 minutes.The drying of the granulated product was carried out in a hot-air oven at 60° C in 4 hours.

Physical measurements of granules- A set of sieves (315,500,800,1250 and 2000 µm) connected to a vibrating apparatus (Erweka AR 400) was used and the particle size distribution was characterized through the determination of the geometric mean diameter by weight (D<sub>50%</sub>,µm) and the geometric standard deviation (r<sub>g</sub>).

The percentage in weight (w/w) of granules which are smaller than 1250  $\mu\text{m}$  is also calculated. Flow rate (g/sec) was carried out by determining the time required to discharge 100 g of granules through a 0.8 cm orifice of a glass funnel.

**Choice of experimental design-** A three-variables, rotatable, central composite design with six center points (Table 1) was employed to generate the data which were then fitted to the second degree polynom (eq.3).

$$Y = b_0 + b_1X_1 + b_2X_2 + b_3X_3 + b_{11}X_1^2 + b_{22}X_2^2 + b_{33}X_3^2 + b_{12}X_1X_2 + b_{13}X_1X_3 + b_{23}X_2X_3 \quad (\text{eq.3})$$

A central composite response surface design was employed because of favorable post experience with such design (6).

## RESULTS AND DISCUSSION

The optimal combination of three independent variables moisture level( $X_1$ ), impeller speed( $X_2$ ), granulation time( $X_3$ ), was sought.

The properties to be optimized and constraint levels were:

geometric mean diameter by weight,	$Y_1 > 200 \mu\text{m}$
percentage of particles smaller than 1250 $\mu\text{m}$	$Y_2 > 80\%$
geometric standard deviation,	$Y_3$ -without constraint
flow rate (g/sec),	$Y_4$ -without constraint

For  $Y_1$  and  $Y_4$  the one-sided transformation, given by eq.2, were used and are shown in Figs 3 and 4. As can be seen, we set  $Y_a=200$ ,  $Y_b=300 \mu\text{m}$  and  $r=0.1$  in the transformation given by (eq.2) for  $Y_1$  and  $Y_a=80\%$ ,  $Y_b=92\%$  and  $r=10$  for  $Y_2$ . While  $Y_3$  and  $Y_4$  were used without constraint and are shown in Figure 5.

The next step was to use the coefficients given in Table 2 along with various values of  $X_1$ ,  $X_2$  and  $X_3$  to obtain the  $Y_i$ . Each  $Y_i$  was then transformed into a  $d_i$



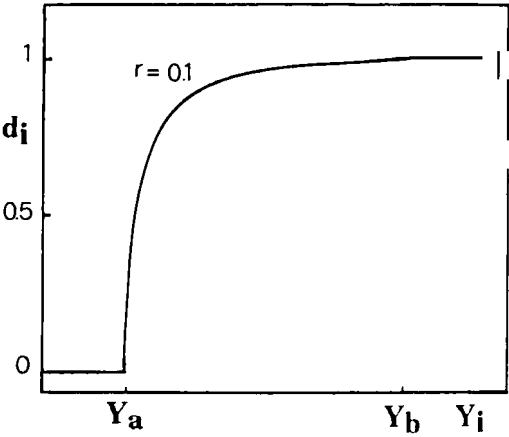


FIGURE 3

Graph of transformation used for  $Y_1$

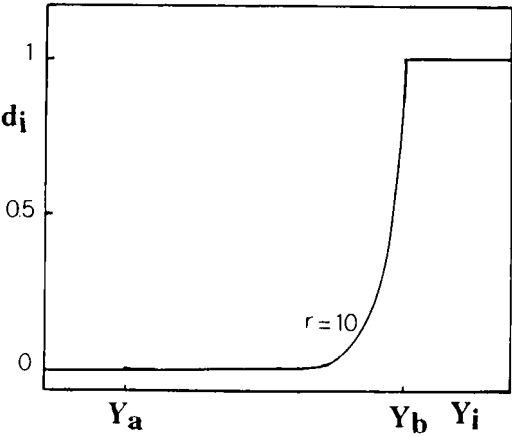


FIGURE 4

Graph of transformation used for  $Y_2$

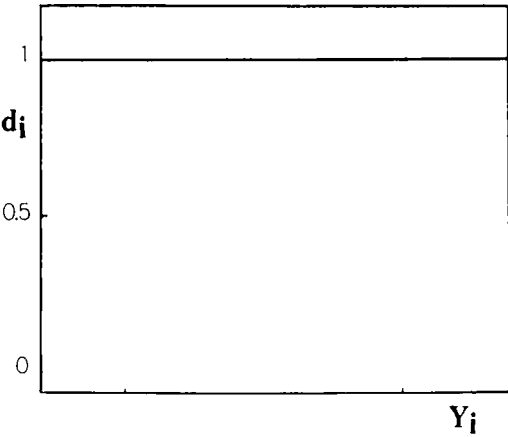


FIGURE 5

Graph of transformation used for  $Y_3$  and  $Y_4$

TABLE 2

Estimated model parameters

	$Y_1$	$Y_2$	$Y_3$	$Y_4$
$b_0$	293.48	53.09	7.83	83.45
$b_1$	53.72	-1.63	-0.89	0.99
$b_2$	-49.76	-6.08	-0.14	7.68
$b_3$	8.18	-0.83	-0.14	1.64
$b_{11}$	20.03	-1.13	-0.13	-0.50
$b_{22}$	59.99	4.82	0.18	-2.48
$b_{33}$	12.61	-0.97	0.31	-0.99
$b_{12}$	21.87	1.47	0.36	-2.74
$b_{13}$	40.87	0.45	-0.05	-1.23
$b_{23}$	18.37	0.70	-0.10	-0.59

TABLE 3

Optimum conditions and predicted properties

Position of the optimal point	
$X_1 = -1.309$ (moisture level, 17.19% w/w)	
$X_2 = 0.943$ (impeller speed, 493 rpm)	
$X_3 = 0.463$ (granulation time, 8 min)	
<u>predicted properties</u>	<u>desirability (d)</u>
$Y_1 = 227$	$d_1 = 0.904$
$Y_2 = 90.74$	$d_2 = 0.360$
$Y_3 = 1.67$	$d_3 = 1$
$Y_4 = 8.33$	$d_4 = 1$
Maximum composite desirability, $D = 0.799$	

using (eq.2) as illustrated in figures 3 and 4. The four  $d_i$  were combined into a single  $D$  using eq.1. Hence, for each level of  $X_1, X_2$ , and  $X_3$ , a  $D$  value was obtained.

We then searched through the levels of  $X_1, X_2$ , and  $X_3$  to find the optimum value for  $D$ . This process was carried out with the program for computer, NEMROD (7). The algorithm employed converged in 3699 iterations.

The resulting optimum formulation is shown in Table 3. The maximum composite desirability was 0.799 and all of the constraints have clearly been met. The value of 0.799 has little physical meaning, except that indicates the level of the  $X$  where the maximum  $D$  occurs.

Figures 6, 7 and 8 show the contour plots of  $D$  for two independent variables with the third one held at its optimum. All these plots present a very sharp surface in the maximum region, so that even a small shift from optimum of the  $X$  values would yield a significant decrease of desirability. (Figs. 9, 10, 11)

The feasibility of scaling-up to 50L high shear mixer was analyzed utilizing the following equation:

$$V_p = V_r P \Pi / 60 \quad \text{eq.4}$$

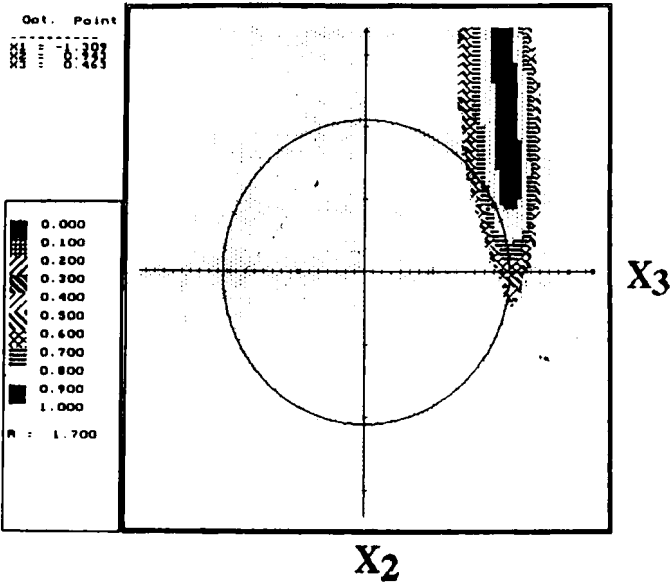


FIGURE 6

Contour plots of D for X<sub>2</sub> and X<sub>3</sub>

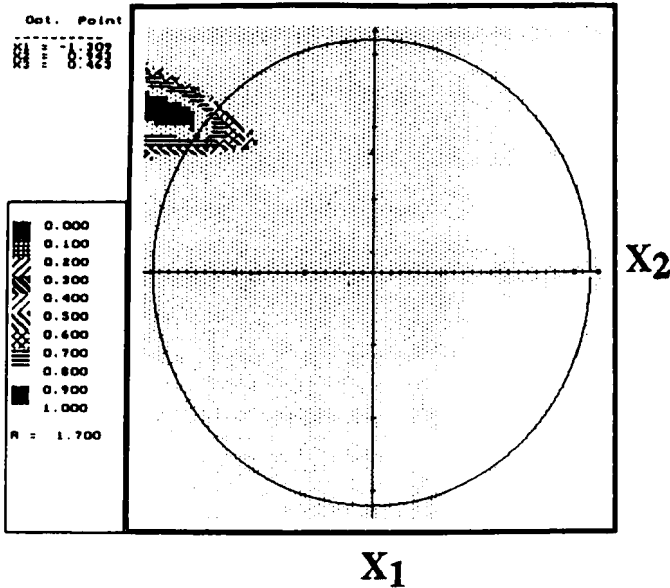


FIGURE 7

Contour plots of D for X<sub>1</sub> and X<sub>2</sub>

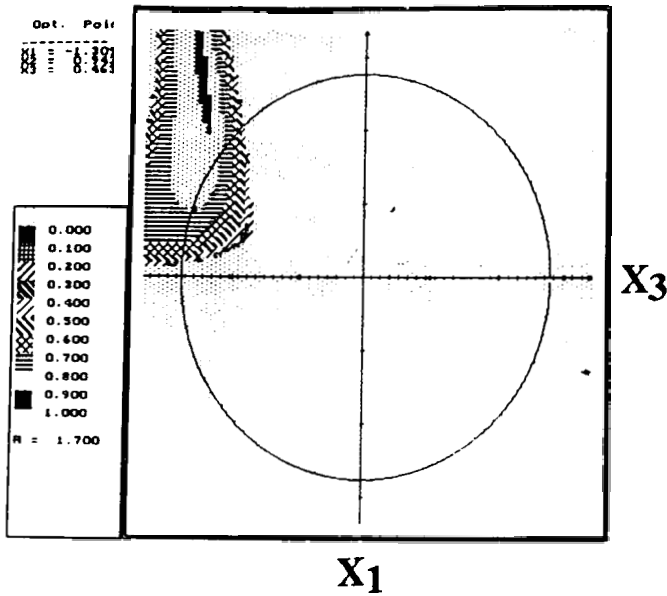


FIGURE 8

Contour plots of D for  $X_1$  and  $X_3$

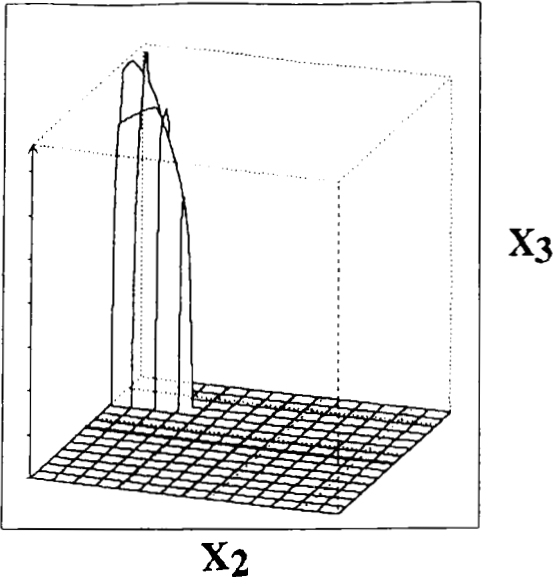


FIGURE 9

Optimum zone for  $X_2$  and  $X_3$ ; and  $X_1 = -1.309$

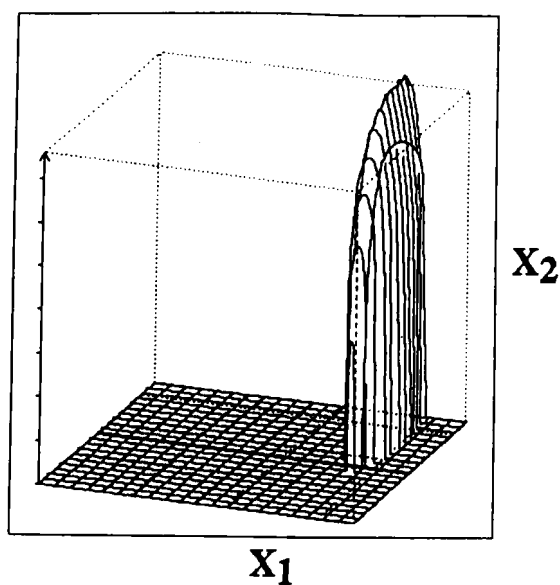


FIGURE 10

Optimum zone for  $X_1$  and  $X_2$ ; and  $X_3 = 0.463$

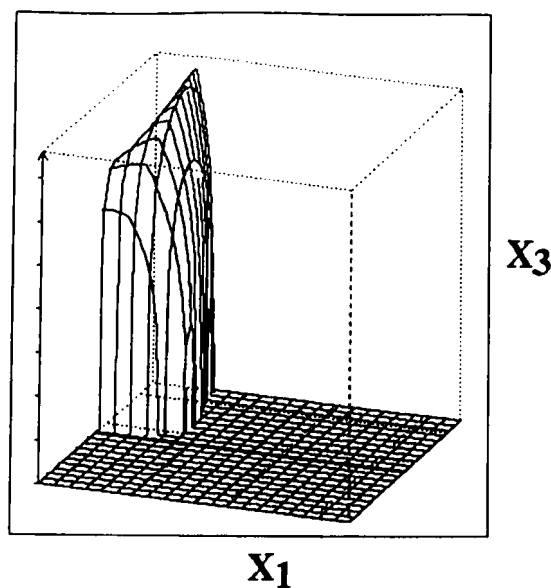


FIGURE 11

Optimum zone for  $X_1$  and  $X_3$ ; and  $X_2 = 0.943$

TABLE 4

Results for "Optimal scaling - up" conditions

Responses	Experimental results		Theory
	10L batch	50L batch	
Geometric mean diameter by weigh( $\mu\text{m}$ )	246	258	227
Geometric standard deviation	3.0	2.5	1.67
percentage of particles smaller than 1250 $\mu\text{m}$ (%)	86	87	90.74
Flow rate (g/s)	7.60	10.52	8.33

where  $V_p$  = periferic rate ( m/ sec )

$V_r$  = impeller rotation speed ( rpm )

$P$  = diameter of the mixer (m)

The results for both batches of different capacity are given in Table 4 where they are compared with the theoretical responses, calculated from equation 3 and from the estimates of the coefficients in Table 2.

The optimum, determined at 10L, proved suitable for scaled up manufacture at 50L.

Data analysis shows a higher value of geometric standard deviation for 10L than 50L batch, as a consequence different flow rates have been obtained due to the different shape of the granules produced (Fig. 12). The mixers used are not

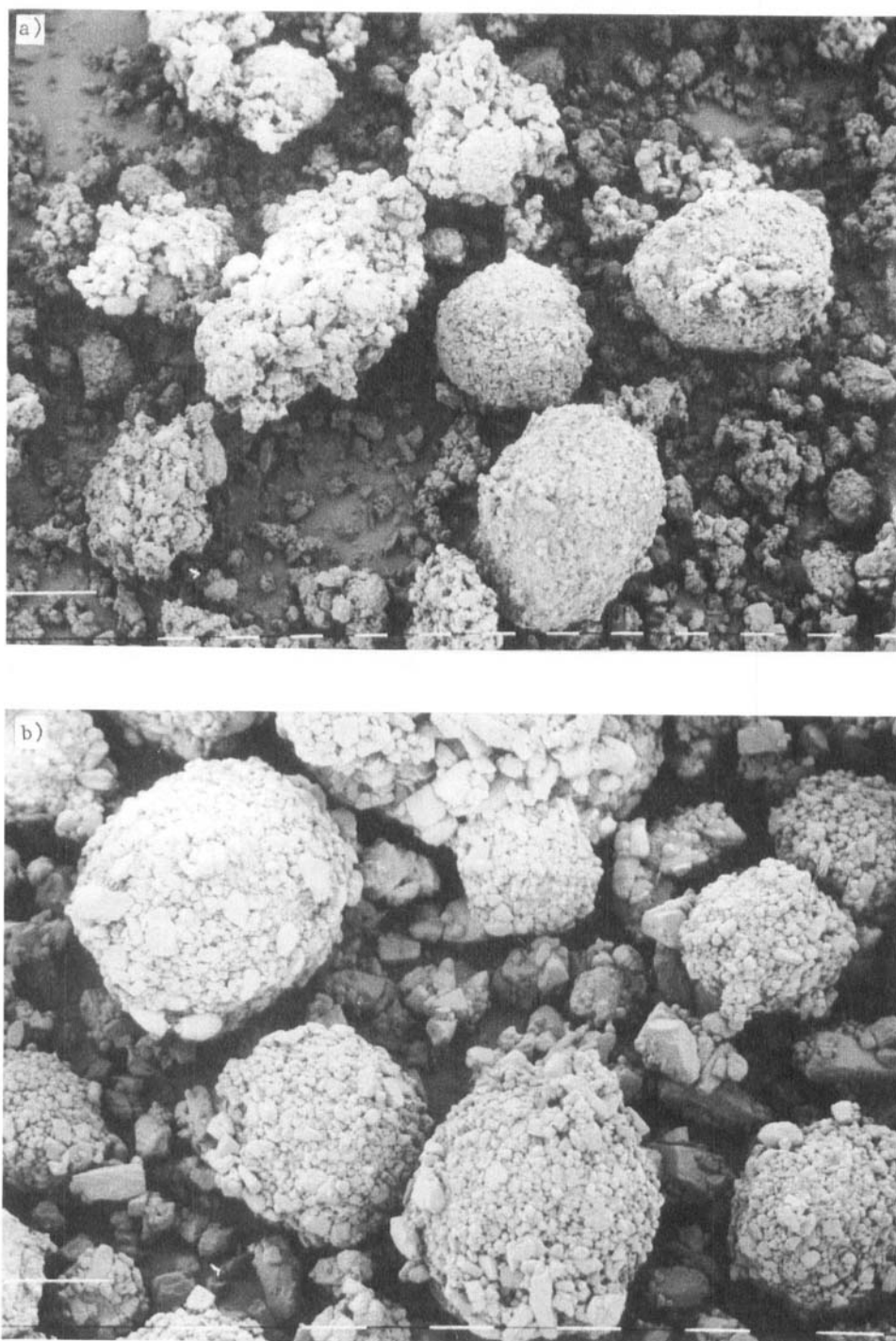


FIGURE 12

SEM photograph of a granule obtained with a: a) Roto J-10L b) Roto P-50L



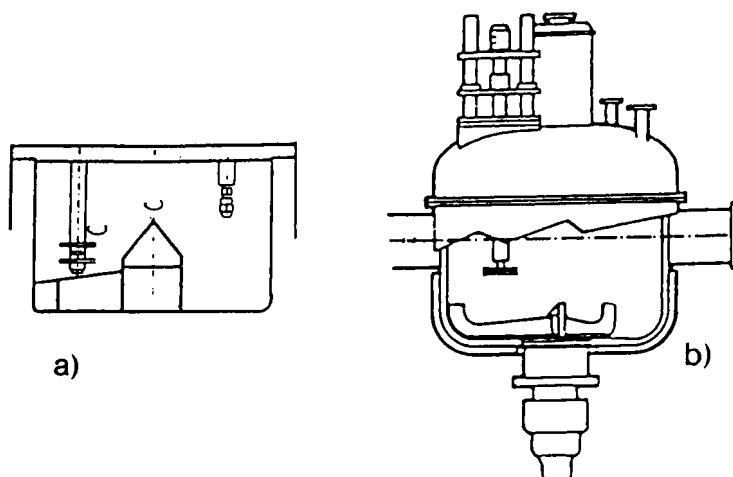


FIGURE 13

Vertical high speed mixer: a) Roto J-10L b) Roto P-50L

geometrically similar and have different internal structure causing (Fig. 13) differences between the granules.

### CONCLUSIONS

Simultaneous optimization of several response variables may be used in the pharmaceutical process of granulation to plan the experimental design and to find optimal zone of the physical properties of the granulate. The method is also recommended for scaling-up study in a high shear mixer.

### REFERENCES

1. N.E. Hartmann and R.A. Beaumont, *Journal of the Institute of the Rubber Industry*, **2**, 272-275 (1968)
2. T.A.J. Nicholson and R.D. Pullen, *Computer Aided Design*, **1**, 39-47 (1969)
3. E.C.Jr. Harrington, *Industrial Quality Control*, **21**, 494-498 (1965)

4. P.E. Gatz and R.C. Millan, Division of Rubber Chemistry, American Chemical Society Fall Meeting, Paper 6, Cincinnati, Ohio, (1972)
5. G. Derringer, R. Suich, Journal of Quality Technology, 12, 214-219, (1980)
6. D. Vojnovic, P. Selenati, F. Rubessa, M. Moneghini, A. Zanchetta, Drug Dev. Ind. Pharm., 9, 961-972 (1992)
7. D. Mathieu and R. Phan-Tan-Luu, Program NEMROD, L.P.R.A.I., Université d'Aix, Marseille, 1990.