

# Hydrodynamic Simulation Behavior of Two Phase Conical Fluidized Column

Abd Ali, K. M.\* , Ghanim, A. N.\*\*

**Abstract**—The use of conical fluidized beds is beginning to receive much attention for biochemical reactions and biological treatment of waste water, also been used successfully in chemical reactions, crystallizations and in other areas . This work was aimed at modeling the fluidization characteristics for a conical fluidized-bed combustor using glass beads as the inert bed material fluidizing in air flow. Response Surface Methodology (RSM) model were used for predicting the pressure drop ( $\Delta p$ ) over the bed length. The predicted results were validated by experimental data of 39 run for other work. Response Surface Methodology (RSM) was applied in the development of modeling, statistical analyzing, and interpreting the resulted correlated variables, response gives the best fitted equation of second-order functions of three variable factors that has been obtained to describe the behavior of pressure drop through the bed. Effects of the glass beads particle size, velocity, and porosity of the bed on the pressure drop and fluidization pattern. The relative computational errors were found to be within 15% with the proposed model.

**Index Terms**— Conical fluidized bed , Trapped bed , Static bed height, RSM, Model, ANOVA.

## 1 INTRODUCTION

Fluidization is the operation by which solid particles behave like a fluid through suspension in a liquid or gas. One of the most important features of fluidized beds is their ability to mix and segregate. Fluidization is the preferred method of operation due to its many advantages over other configurations, like; good solid mixing leading to uniform temperature throughout the bed, high mass and heat transfer rates, easy solids handling, ability to maintain a uniform temperature, significantly lower pressure drops which reduce pumping costs, lower investments for the same feed and product specifications, yielding large axial dispersion of phases, etc, (Fan L. S. 1989)

In fluidization operation, fine solids are transformed into a fluid like state through contact with either a gas, liquid or both. Under the fluidized state, the gravitational pull on granular solid particles is offset by the fluid drag on them, thus the particles remain in a semi-suspended condition. At the critical value of fluid velocity the upward drag forces exerted by the fluid on the solid particles become exactly equal the downward gravitational forces, causing the particles to become suspended within the fluid. At this critical value, the bed is said to be fluidized and exhibit fluidic behavior (Shailendra, 2010). Gas-solid fluidized bed reactors have found a wide range of industrial applications The use of fluidized beds make possible to obtain a great homogeneity of distribution of the solid phase and a great effectiveness of mass and heat exchanges. In fact, the division of the solid allows having a great surface available for the exchanges and particles agitation accelerates

the transfer processes(Davidson et al, 1985).

Conical fluidized bed is very much useful for the fluidization of wide distribution of particles, since the cross sectional area is enlarged along the bed height from the bottom to the top, therefore the velocity of the fluidizing medium is relatively high at the bottom, ensuring fluidization of the large particles and relatively low at the top, preventing entrainment of the small particles. Since the velocity of fluidizing medium at the bottom is fairly high, this gives rise to low particle concentration, thus resulting in low reaction rate and reduced rate of heat release. Therefore the generation of high temperature zone near the distributor can be prevented (Biswal K. C. et al. 2010). Due to the existence of a gas velocity gradient along the height of a conical bed, it has some favorable special hydrodynamic characteristics. The conical bed has been widely applied in many industrial processes such as, biological treatment of wastewater, immobilization biofilm reaction, incineration of waste-materials, fluidization of cohesive powder, coating of nuclear fuel particles, crystallization, roasting of sulphide ores, coal gasification and liquefaction, etc (Rachadaporn et al. 2006) (see Fig.1).

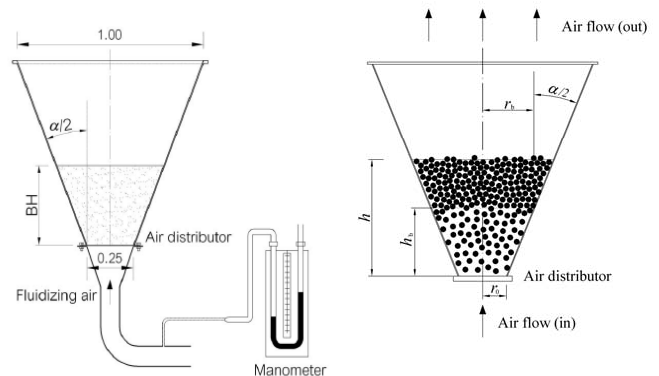


Fig. 1 Schematic of the conical bed with solid particles fluidized by air flow for partially fluidized bed mode. (Rachadaporn et al. 2006).

\* Abd Ali, K. M., PhD, is currently Asst. Professor in Electrochemical Engineering Department, Engineering College, Babylon University, Hilla 51002, PO box 4, Iraq, E-mail: [alassade\\_67@yahoo.com](mailto:alassade_67@yahoo.com)

\*\* Ghanim, A. N. , Corresponding Author, is currently Asst. Professor in Electrochemical Engineering Department, Engineering College, Babylon University, Hilla 51002, PO box 4, Iraq, E-mail: [ala\\_gh2003@yahoo.com](mailto:ala_gh2003@yahoo.com)

The main objective of present work is to apply the response surface methodology (RSM) to accomplish relationships between pressure drop and variables such as velocity, particle size and porosity using Minitab software (version 15) and to model conical fluidized bed behavior, as well examine the validity of model results due to literature experimental data (Biswal K. C. et al. 2010).

## 2 Modeling of Experimental Data

Response surface methodology is a collection of mathematical and statistical techniques useful for the modeling and analysis of problems in which a response of interest is influenced by several variables and the objective is to optimize this response (Montgomery, 2005; Khuri and Mukhopadhyay, 2010), the present RSM procedure was carried out as follows:

- 1) A series of 39 experiments were designed of reliable measurement for conical fluidized bed with packing of glass beads.
- 2) Mathematical model of the second-order response surface were developed.
- 3) The set of experimental factor variables producing the optimal response value were determined.

A quadratic regression model was employed for predicting the optimum conditions. Each response of Y can be represented by a mathematical equation that correlates the response surface. The response (Y) can be expressed as polynomial model (Montgomery, 2005; Khuri and Mukhopadhyay, 2010) according to Eq. (1):

$$Y = \beta_0 + \sum_{i=1}^k \beta_i x_i + \sum_{i=1}^k \sum_{j=i+1}^k \beta_{ij} x_i x_j + \sum_{i=1}^k \beta_{ii} x_i^2 \quad (1)$$

Where Y is the predicted response used as a dependent variable k is the number of independent factors, xi (i = 1, 2) the controlling factors;  $\beta_0$  the constant coefficient, and  $\beta_i$ ,  $\beta_{ij}$  and  $\beta_{ii}$  the coefficients of linear, interaction and quadratic term, respectively.

Analysis of variance (ANOVA) was used for graphical analysis of data to obtain the interaction between factors and response. The quality of the fit polynomial model was expressed by the regression coefficients  $R^2$  and its statistical significance was checked by the Fisher's F-test (Montgomery, 2005). Model terms were evaluated by the p-value (probability) with 95% confidence level. The coefficient parameters were estimated using response surface regression analysis employing the Minitab software version 15, also used to find the residuals, 3D surface and 2D contour plots of the response models.

## 3 MATERIALS AND METHOD

The present work is concentrated on an understanding the complex hydrodynamics of two phase fluidized bed. Fluidized bed of height 1.88 m and diameter 0.1 m (Moharana and Malik, 2010) have been simulated. The solid phase used is glass beads of size 2.18 mm in the present work. Co-current gas-solid fluidization with gas as continuous phase has been used. The static bed heights of the solid phase in the fluidized bed used for simulation are taken as 21.3 cm. Initial solid hold up has been taken as 0.59 in all cases with superficial velocity of gas varying in the range of 0.025-0.1 m/sec. Table (1) show

the properties of simulated model.

TABLE (1) Properties and characteristics of glass bed and air (Biswal K.C et al. 2010).

Property/characteristic	Value
Glass particle size	0.00218- 0.00258 m
Glass beads porosity	0.385- 0.441
Density of glass beads, $\rho_s$	2300 kg/m <sup>3</sup>
Sphericity of glass beads particles, $\phi_s$	0.88
Density of air, $\rho_f$	1.165 kg/m <sup>3</sup>
Viscosity of air, $\mu_f$	1.86x10 <sup>-5</sup> N.s/m <sup>2</sup>

## 4 Results and discussion

The experimental data were adjusted by means of a regression method in order to describe the pressure drop as a polynomial function of air flow, particle size, and porosity.

### 4.1 Fluidization Characteristics

Fig. 2 shows the interaction effect of fluidization with air velocity, particle size, and porosity of glass bead particles on pressure drop of bed. Interactions plot creates a single interaction plot for three factors. An interaction plot is a plot of means for each level of a factor with the level of a second factor held constant. Interaction is present when the response at a factor level depends upon the levels of other factors. The greater the departure of the lines from the parallel state, the higher the degree of interaction. Consequently, the interaction effect of the above mentioned three factors on pressure drop shows that there are significant interactions between factors on the response as confirmed with response equation.

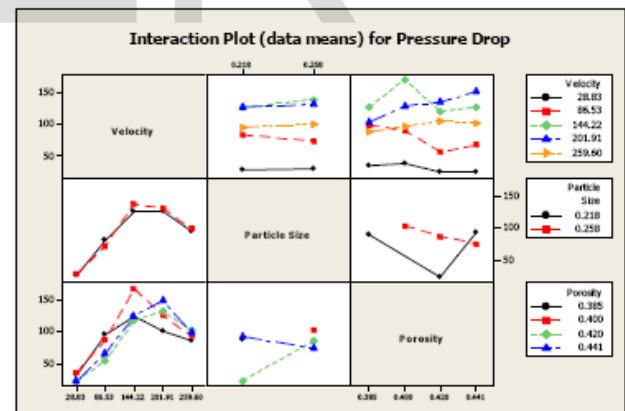


Fig.(2) Interaction effect among the parameters pressure drop, velocity, particle size, and porosity

### 4.2 Statistical Analysis

The relationship among the three factor variables and the pressure drop response for fluidization process was analyzed using response surface methodology (RSM). The Multilevel Factorial Design (MFD) shown in Table (2) allowed the development of mathematical equation where predicted response Y was assessed as a function of velocity, particle size and porosity with number of levels taken to be 5, 2, and 4, respectively.

This study involves 39 runs of experiment response. The

MFD reduces the number of experimental trials, however 23 runs were performed and all experimental results were treated and interpreted by statistical analysis.

TABLE (2) Data Matrix and Multilevel Factorial Design Results

un No.	Velocity (cm/s)	Particle size (cm)	porosity	Pressure drop ( $\Delta P/l$ ) (Pa/cm)	
				Exp. *	Pred.
1	201.9	0.258	0.42	134.41	133.83
2	28.8	0.218	0.42	24.72	27.103
3	144.2	0.258	0.441	126.5	127.39
4	201.9	0.258	0.4	128.28	132.46
5	144.2	0.218	0.385	127.04	123.37
6	201.9	0.218	0.441	150.92	125.87
7	86.5	0.218	0.385	98.18	94.18
8	28.8	0.218	0.441	24.39	23.21
9	28.8	0.218	0.385	33.59	33.52
10	144.2	0.258	0.4	169.66	128.32
11	259.6	0.258	0.42	105.05	108.32
12	259.6	0.218	0.385	88.47	87.36
13	259.6	0.218	0.441	101.26	97.17
14	201.9	0.218	0.385	102.15	121.10
15	86.5	0.218	0.441	67.07	88.90
16	259.6	0.258	0.4	95.62	105.15
17	86.5	0.258	0.42	55.62	90.46
18	86.5	0.258	0.4	89.83	92.70
19	144.2	0.258	0.441	126.53	123.12
20	28.8	0.258	0.441	24.39	17.32
21	144.2	0.258	0.42	119.48	127.88
22	28.8	0.258	0.4	38.05	25.61
23	28.8	0.258	0.42	24.72	21.58

\* Biswal K.C. et al. 2010

### 4.3 Simulation Model

Partial least squares PLS fits multiple response variables in a single model. Because PLS model the response in a multivariate way, the results may differ significantly from those calculated for the response individually. Observed responses were utilized to conduct models using response surface methodology (RSM) (Montgomery, 2005). To develop a response surface regression model, a general polynomial model (Eq. 6) was applied to experimental observations of the pressure drop response (Y), and quadratic regression models were obtained. The predicted regression equation is:

$$\frac{\Delta P}{L} = 76.018 + 0.517v - 9.708D_p - 101.97\epsilon - 0.005v^2 - 20.396D_p^2 - 39.479\epsilon^2 + 2.202vD_p - 433.746D_p\epsilon + 1.557v\epsilon \quad (2)$$

Where:  $v$  is velocity factor,  $D_p$  is particle size factor, and  $\epsilon$  is porosity factor. The accuracy of equation with  $R^2$  equal 0.888.

### 4.4 Analysis of Variance

The results obtained were then analyzed by ANOVA as shown in table (3), to assess the "goodness of fit". The ANOVA analyzed has been tested for full quadratic equation designed value and gives an insight into the linear, quadratic and interaction effects of the factors. The p-value is used as a tool to check the significance of each factor. It was found that the fac-

tors with major effect on pressure drop were the linear effect of particle size of packing and its porosity having p-values of 0.000, followed by velocity of gas phase and porosity quadratic factors having p-values 0.004 and 0.016, respectively, and latest the particle size against porosity interaction factor, with p-value of 0.0972.

TABLE (3) Analysis of Variance (ANOVA)

Source	DF	SS	Mean SS	F-value	P-value
Regression	4	38852	9713	35.93	0.000
Residual Error	18	4866.6	270.37		
Total	22	43718			

Analysis of variance results of quadratic models presented in Table (3) indicated the reduced quadratic models statistical parameters of degree of freedom DF, sum of squares SS due to regression and residual error, mean sum of squares MSS due to regression and residual error, value of Fisher's test, and regression p-value.

The p-values of ANOVA Table (3), indicates that the relationship between responses and factors is statistically significant at a level of  $\alpha = 0.05$  (maximum acceptable level of risk for rejecting a true null hypothesis, Montgomery, 2005).

### 4.5 Validation of Model

The response surface model were developed in this study with values of  $R^2$  higher than 88%. It is usually necessary to check the fitted model to ensure it provides an adequate approximation to the real system. The  $R^2$  coefficient gives the proportion of the total variation in the response predicated by the model, indicating ratio of sum of squares due to regression (SSR) to total sum of squares (SST) (Dean and Voss, 1999; Montgomery, 2005).

Graphical method usually used to validate models, and also characterizes the nature of residuals of the model. A residual is defined as the difference between an observed value Y and its fitted  $\hat{Y}$  (Montgomery, 2005). In the normal probability plot, this was used to check the normality distribution of the residuals as shown in Figure(3).

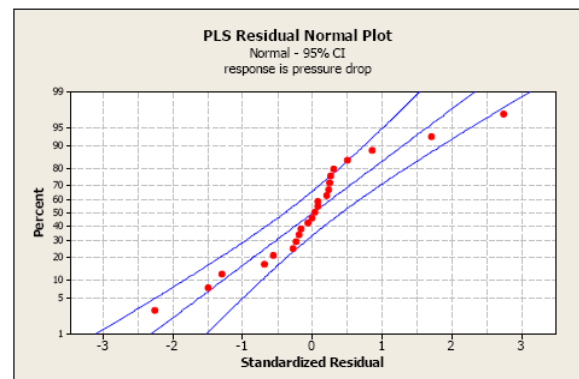


Fig.3 Normal Probability of pressure drop Response

The error distribution may be slightly skewed, with the right and left tails. Usually, the normal probability plot has a

tendency to bend up or down slightly on the left or right sides implies that the error distribution is somewhat thinner than would be anticipated in a normal distribution (Montgomery, 2005); that is, the negative residuals are not quite as large (in absolute value) as expected.

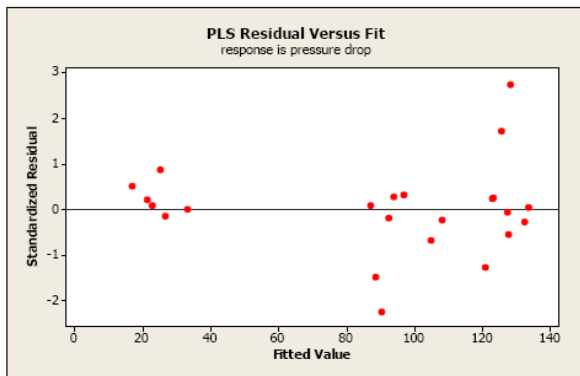


Fig.4 Residuals vs. Fits of pressure drop

If the model is correct and if the assumptions are satisfied, the residuals should be structureless; they should be unrelated to any other variable including the predicted response. A simple check is to plot the residuals versus the fitted values  $\hat{Y}$  (Montgomery, 2005); Accordingly, Figure (4) plot not reveal any obvious pattern.

Fig.5 shows the optimal design(a group of the best design points selected when reducing or augmenting the number of experimental runs in the original design). By taking the number of components cross-validated equal to 8, the D-optimality minimizes the variance in the regression coefficients of the fitted design model, thereby providing the most precise estimate of the effects.

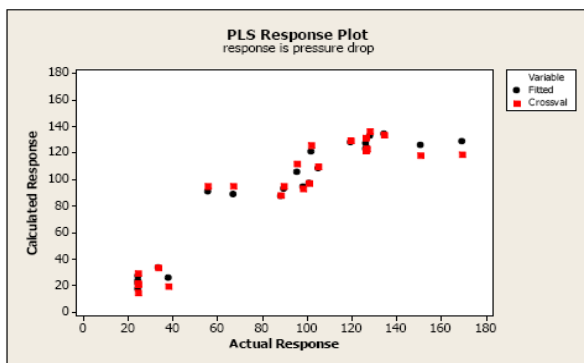


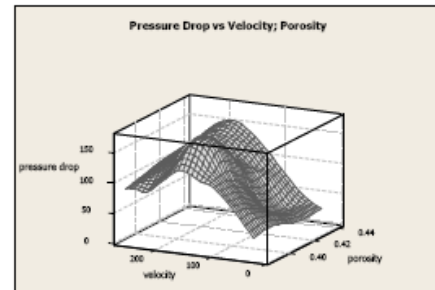
Fig.5 Actual and calculated response

The predicted pressure drop corresponding to experimental values for glass bead particles of size within 0.00218–0.00258 m, for different static bed heights was acceptable.

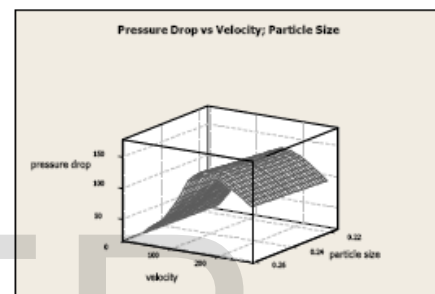
However, the computational and experimental values were in rather good agreement (within 15% relative error) for 5.6–10 cm static bed heights, regardless of the cone angle. For operating conditions applied in the tests with these particle size range, the predicted  $\Delta p_{mf}$  as shown in Figure (5) were found to be lower than corresponding experimental values.

#### 4.6 Surface and Contour Analysis

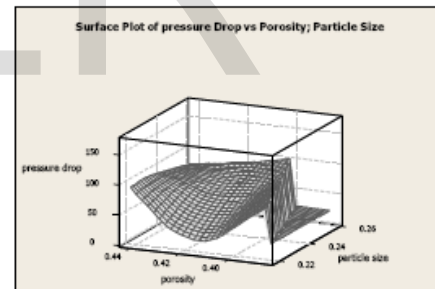
The result indicates that the linear and quadratic factors were significant in determining the response value of pressure drop along static head for the fully fluidization mode and also significant in the statistical analyses.



(a)



(b)



(c)

Fig.(6) Surface Interaction effect among the parameters pressure drop , velocity , particle size, and porosity.

Moreover, the analysis implies that the model were definitely influenced by the three factors. The response surface and contour plots are the graphical representation of the regression equation used to visualize the relationship between the response and experimental levels of each factor.

The response surface plots can be observed in Figure (6) that shows 3D response surface plots for the three variable factors and all curves showing very noticeable changes. The resulted data of conical fluidized bed  $\Delta p_{mf}/l$  was represented by irregular surface with highest plate zone approaches center of plate for the cases of (6a) and (6b).

This trend was clearly viewed for pressure drop as a function of air velocity and porosity factors as well as air velocity

and particle size. In contrast, an explicit observation of pressure drop is shown in (6c), implying that the optimum conditions for the predicted  $\Delta p_{mf}/l$  against corresponding experimental values for glass bead particles of medium size, 218–258  $\mu\text{m}$ , for different static bed heights and 10.37° cone angles.

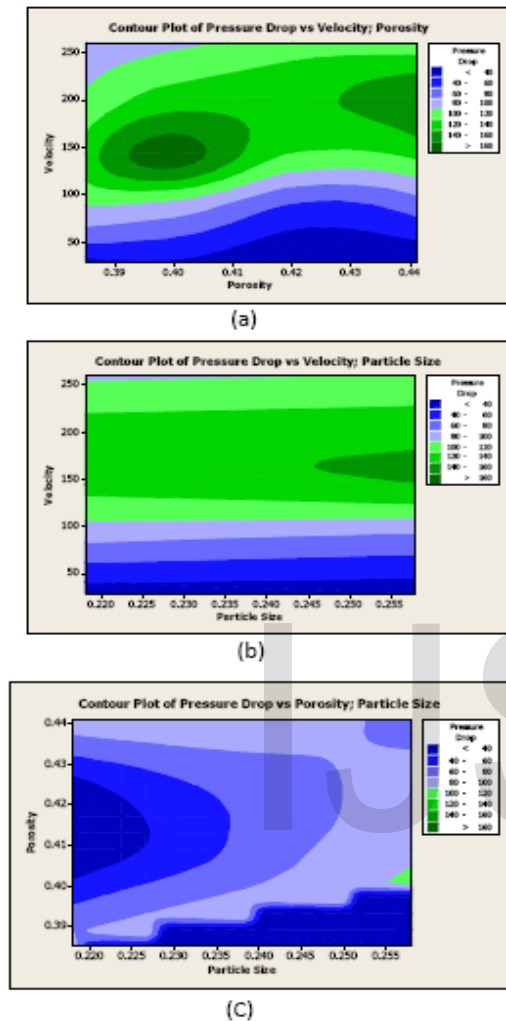


Fig.(7) Contour interaction effect among the parameters pressure drop , velocity , particle size, and porosity.

Contour plots of pressure drop are shown in Figures (7a), (7b), and (7c) with 2D. In a contour plot, the values for three factor variables for a response variable is represented by homogeneous shaded zones, called contours. These plots present the overall distribution of pressure drop along the length of the bed . As shown, increased pressure drop was observed with increasing velocity until minimum fluidization condition take place and then will be decreasing when the bed will be spouted.

## 5 CONCLUTIONS

Response Surface Methodology (RSM)was used to study the hydrodynamics of gas-solid fluidized bed. Response Surface Methodology (RSM) was successfully applied for predicting the fluidization characteristics, the minimum fluidization velocity as well as the dependence of the pressure drop over

the bed on the superficial air of the bed for conical air –glass beads fluidized bed. The effects of operating variables (velocity, particle size and porosity) on pressure drop were found to be significant. The model have been validated by comparison of the predicted results with experimental data for specific range of the above operating conditions.

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