



# Statistically Designed Bioprocess for Enhanced Production of Alkaline Protease in Bacillus cereus HP\_RZ17

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Alkaline protease is one of the bulk enzymes having wide commercial demand for various applications. It is commercially produced by a submerged fermentation process employing various bacteria, *Bacillus* sp. being the most widely used species. Statistical optimization of the process for the production of alkaline proteases from rhizospheric bacteria and its application in the biocontrol of plant pathogens has not been explored fully and needs to be studied for the development of efficient bioprocess. We report the enhanced production of alkaline protease in the minimal salt medium (MSM) optimized using statistical approaches such as Plackett Burman Design (PBD) and Response Surface Methodology (RSM). In the first step; PBD, among the total eight variables, three variables namely, yeast extract (p<0.05), fructose (p<0.05) and pH (p<0.05) influenced the production of alkaline protease by *Bacillus cereus* HP\_RZ17. These three variables were further analyzed in the second step i.e. Central Composite Design (CCD) of RSM. The optimum yield of alkaline protease by *B. cereus* HP\_RZ17 (130.72 UmL<sup>-1</sup>) was obtained under the optimal conditions such as yeast extract (0.899% w/v), fructose (0.873% w/v), and pH (11.25) of production media. The statistically optimized values of variables used for the scale-up of the process at 5 L capacity bioreactor enhanced the alkaline protease yield (132.48 UmL<sup>-1</sup>) by 1.09 fold vis-à-vis un-optimized protocol (121.96 UmL<sup>-1</sup>) in *B. cereus* HP\_RZ17.

Keywords: Central Composite Design, Optimization, Plackett Burman Design, Response Surface Methodology, Scale-up

# Introduction

Alkaline proteases [EC.3.4.21-24, 99] are commercially versatile enzymes capturing about 45% of the total global enzyme market.<sup>1</sup> Owing to the broad range of working temperature (10–80°C) and pH  $(4-12)^{2-6}$ , microbial protease; especially those produced by *Bacillus* sp. have a huge demand in the enzyme market while protease produced by rhizobacteria have been extensively used in agriculture for the biocontrol of a wide variety of phytopathogens.<sup>7</sup>

Members of the genus *Bacillus* constitute an important group of plant growth promoting rhizobacteria (PGPR).<sup>8,9</sup> Genus *Bacillus* constitutes one of the largest easily cultivable, fastidious groups of PGPR which improve growth and yield of crops.<sup>9</sup> Bacilli, due to their strong rhizosphere colonization, ability to produce very hard spores which survive for prolonged periods in soil<sup>10</sup> and salinity tolerance have been seen as effective bio-

inoculants for plant growth promotion.<sup>11</sup> Members of *Bacillus* species successfully colonize the rhizospheric regions<sup>12</sup> and contribute to better growth and yield of crop plants. They have been comprehensively studied due to their effective plant growth promoting ability and antagonistic activity against wide range of phytopathogens.<sup>13</sup> *Bacillus* sp. promote plant growth through mineral nutrition, nitrogen fixation and production of phytohormones and hydrolytic enzymes.<sup>14</sup>

Statistical based optimization offers the best alternative solution to overcome these demerits of OVAT.<sup>15–18</sup> The present research work was focused on the optimization of different variables of production media based on statistical approaches, such as PBD and RSM by CCD to escalate the production of alkaline protease in *B. cereus* HP\_RZ17.

# **Materials and Methods**

#### Source of culture

The isolate *Bacillus cereus* HP\_RZ17 used in the present study was obtained from culture repository of

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Department of Microbiology of PSGVPM's ASC College, Shahada, India.<sup>19</sup>

# Evaluation of medium variables by OVAT for the production of alkaline protease

The influence of various physicochemical factors like incubation period (0-96 h) and temperature media pH (4.00-12.00), inoculum (20–50°C), concentration (0.5-2.0% v/v), nitrogen (1.0% w/v)(casein, tryptone, veast extract, urea, ammonium nitrate (NH<sub>4</sub>NO<sub>3</sub>) and ammonium chloride (NH<sub>4</sub>Cl<sub>2</sub>) and carbon sources (1.0% w/v) (glucose, fructose, sucrose, lactose, maltose, and dextrose) on the production of alkaline protease in B. cereus HP RZ17 were initially studied individually while keeping the other constituents constant using the OVAT methodology. The alkaline protease production was carried out in MSM containing  $(gL^{-1})$ ,  $KH_2PO_4$ ; 0.002,  $MgSO_4$ ; 0.005, NaCl; 0.005, CaCl<sub>2</sub>, 0.005, yeast extract; 10, fructose; 10, pH 11.<sup>20,21</sup> The lag phase batch culture of B. cereus HP\_RZ17 (1.5% v/v) was grown in MSM at  $30^{\circ}$ C in shaking at 120 rpm for 48 h.<sup>22,23</sup> Following the incubation, the broth was centrifuged at 10,000 rpm for 15 min at 30°C and the resulting supernatant was subjected to protease assay.<sup>19</sup>

#### **Protease assay**

Protease activity was estimated in terms of proteolysis of a substrate (casein) and the release of tyrosine units. The cell-free supernatant (crude enzyme) and 0.65% (w/v) casein were incubated in 20 mM phosphate buffer pH 7.0 for 10 min at 37°C, followed by the addition of 110 mM Tri-chloroacetic acid (TCA) to stop the enzyme reaction. Cell-free supernatant was obtained following the centrifugation (10,000 rpm for 10 min) of precipitated protein, was subjected for the protease assay and the number of tyrosine units produced due to proteolysis were estimated.<sup>24</sup> The total protein content was measured as per the method of Lowry *et al.*<sup>25</sup> using Bovine Serum

Albumin (BSA) as a standard. One unit of alkaline protease activity was defined as the amount of the enzyme that liberated 1  $\mu$ g tyrosine mL<sup>-1</sup> min<sup>-1</sup> from casein under specified assay conditions.<sup>19,26</sup>

# Statistical-based optimization of production medium

Identification of significant media components that influence the production of alkaline protease is the most crucial step in the optimization of parameters.<sup>27,28</sup> The significant variable(s) were optimized in the first step i.e. PBD while in second step RSM (through CCD) concentration of significant variable(s) was analyzed for the optimum production of alkaline protease by *B. cereus* HP\_RZ17.<sup>29,30</sup>

# Evaluation of significant variable using PBD

To investigate the significant variables for the optimum production of alkaline protease, eight variables were selected in PBD. This design was based on the two-factorial or multi-factorial experiments, which facilitate to ascertain that the variables have a significant role in the production of an enzyme.<sup>31</sup> The physicochemical parameters that are most likely to affect alkaline protease production preferred were  $K_2HPO_4$  (A), MgSo<sub>4</sub>·7H<sub>2</sub>O (B), CaCl<sub>2</sub> (C), NaCl (D), yeast extract (E), fructose (F), inoculum concentration (G) and pH (H). Each independent variable was examined at two different levels i.e. low (-1) and high (+1).

The PBD experimental design comprised of 12 sets of different levels of eight variables. The effects of all variables on alkaline protease production were calculated and analyzed by one-way Analysis of Variance (ANOVA). Based on the *p*-value (p<0.05) and the highest confidence level <95% of significant variables were selected from other independent variables (Table 1). The magnitude of the selected eight variables was estimated using the Pareto chart analysis (Fig. 1).<sup>32,33</sup>

Table 1 — Experimental PBD data analysis using ANOVA							
Component code	Components	Effect	coefficient	<i>t</i> - value	<i>p</i> -value		
Constant			105.95	101.17	0		
А	$K_2HPO_4(\%)$	-4.30	-2.15	-2.05	0.132		
В	$MgSO_4(\%)$	-4.22	-2.11	-2.01	0.137		
С	$CaCl_2(\%)$	-1.21	-0.60	-0.58	0.605		
D	NaCl (%)	-5.32	-2.66	-2.54	0.085		
Е	Yeast extract (%)	19.80	9.90	9.46	< 0.003		
F	Fructose (%)	12.61	6.30	6.02	< 0.009		
G	Inoculum (%)	4.71	2.36	2.25	0.110		
Н	pН	20.37	10.19	9.73	< 0.002		



Fig. 1 — Pareto chart representing the effects of medium variables on alkaline protease production by *B. cereus* HP\_RZ17

Optimization of significant medium variables by CCD of RSM

The significant variables of production medium obtained from PBD were analyzed to find the optimum region, type of response by variable in that and the data validation using CCD<sup>34</sup> of RSM.<sup>18</sup> The optimized values of variables were illustrated by considering the fractional factorial, distance  $\alpha$  from its center and a central point. In CCD the interactions between significant variables were studied by which the optimal level of the variable can be decided. Among the eight variables identified in PBD yeast extract (E), fructose (F) and pH (H) were identified as significant variables. Hence these variables were picked up for further optimization at five different levels i.e.  $-\alpha$ , -1, 0, +1 and  $+\alpha$ . The range of extremely low ( $\alpha$ ) to extreme high (+ $\alpha$ ) values was considered in CCD. The twenty different experimental sets comprising of eight factorial points, six axial points and central points (Table 2) were experimentally performed and the effects of independent variable or response were calculated in the form of yield (Y) and were expressed as UmL<sup>-1</sup>.

The experimental results of RSM were subjected to ANOVA and fitted into the following second-order polynomial equation (Eq. 1). It generated an empirical model that relates to the responses obtained in the independent variable to the experiment. A secondorder polynomial equation (Eq. 1) was then fitted to the response by multiple regression procedures. This resulted in an empirical model that related the response measured in the independent variables to the experiment.

Table 2 — Experimental data obtained for significant							
variables obtained from PBD in CCD							
Run	Туре	Е	F	Н	Enzyme yield (UmL <sup>-1</sup> )		
					Experimental	Predicted	
1	Factorial	-1	-1	-1	90.52	90.45	
2	Factorial	1	-1	-1	100.27	106.38	
3	Factorial	-1	1	-1	83.64	86.77	
4	Factorial	1	1	-1	109.56	114.51	
5	Factorial	-1	-1	1	96.75	94.55	
6	Factorial	1	-1	1	99.86	99.48	
7	Factorial	-1	1	1	112.67	109.31	
8	Factorial	1	1	1	123.24	126.06	
9	Axial	-1.6818	0	0	90.55	93.36	
10	Axial	1.68179	0	0	127.54	120.84	
11	Axial	0	-1.6818	0	95.61	94.88	
12	Axial	0	1.68179	0	117.29	114.13	
13	Axial	0	0	-1.6818	103.57	96.50	
14	Axial	0	0	1.68179	106.48	109.66	
15	Central	0	0	0	128.97	128.93	
16	Central	0	0	0	128.73	128.93	
17	Central	0	0	0	128.24	128.93	
18	Central	0	0	0	128.45	128.93	
19	Central	0	0	0	129.34	128.93	
20	Central	0	0	0	129.18	128.93	

$$Yi = \beta_0 + \sum \beta_i X_i + \sum \beta_{ii} X_i^2 + \sum \beta_{ij} X_i X_j \qquad \dots (1)$$

Where

Yi - predicted response,

 $X_iX_j$  - input variables (influence the response variable Y),  $\beta_0$  - constant,

- $\beta_i$  i<sup>th</sup> linear coefficient,
- $\beta_{ii}$  quadratic coefficient,
- $\beta_{ij}$  ij<sup>th</sup> interaction coefficient.

The experimental run with the optimized level of the significant medium variables was performed to validate the predicted model.

### Scale-up at laboratory scale (5L) bioreactor

For the validation of the process designed at shake flask level production of alkaline protease by *B. cereus* HP\_RZ17, the protocol was scaled-up to 5L capacity on a fully-automated laboratory bioreactor [Model LF-5 Murhopye Scientific Co., Mysore, India].

# Software and data analysis

Statistical and graphical analysis of experimental data of PBD and RSM was performed with the software Minitab 18 (Minitab GmbH, Munich, Germany).

# **Results and Discussion**

# Evaluation of medium variables by OVAT for the production of alkaline protease

The effective levels of physicochemical i.e. incubation period 48 h, incubation temperature 30°C, media pH 11.0, inoculum concentration 1.5%, carbon source (fructose) 1% and nitrogen source (yeast extract) 1% on optimum alkaline protease production were studied by OVAT methodology<sup>19</sup>, the optimum of an enzyme in MSM production by B. cereus HP RZ17 was achieved up to 121.96 UmL<sup>-1</sup>. Chandra *et al.*<sup>35</sup> reported the influence of various media parameters on protease production by Bacillus sp. (GU566359) evaluated using the OVAT approach where they achieved 470 UmL<sup>-1</sup> of enzyme production. Polley and Ghosh<sup>36</sup> reported the influence of media components on alkaline protease yield in A. alternata TUSGF1 and have claimed an optimum enzyme yield of 96 UmL<sup>-1</sup>.

### Evaluation of significant variable using PBD

The limitation of OVAT to exploit the effect of interaction between media components becomes the major factor to obtain maximum yield in scale-up. To overcome this issue the PBD and RSM have been found as the best advances in the field of media optimization.<sup>37</sup> Using the PBD the most significant and influential variables were identified for the maximum alkaline protease production from *B. cereus* HP RZ17. The standardized effects of eight independent variables analyzed by ANOVA are graphically represented in the Pareto chart (Fig. 1). The yield of alkaline protease produced from B. cereus HP RZ17 in 12 sets of the experiment proposed by PBD estimated under standard conditions was observed in the range of 87.59 - 130.65 UmL<sup>-1</sup>. Analysis of regression coefficients and t -the value of eight variables of the proposed PBD showed that among the tested variables yeast extract (E) (p=0.003), fructose (F) (p=0.009) and pH (H) (p=0.002) positive effects on alkaline protease production. These three variables were found significant for alkaline protease production at a 95% confidence level ( $\alpha = 0.05$ ) (Fig. 1) and hence these variables were chosen for further optimization by RSM.

# Optimization of significant medium variables by CCD of RSM

In the initial screening of variables by PBD, three variables namely E, F, and H came out as significant variables for the production of alkaline protease by *B. cereus* HP\_RZ17. The experimental results of CCD for alkaline protease production showed a good fit with the predicted values (Table 2).

In RSM the variables of Eq. (1) were evaluated using multiple regression analysis. The second-order polynomial regression equation showed the empirical relationship between the alkaline protease yield (Y) and the significant variables of PBD in coded units, the yield was calculated by Eq. (2).

 $\begin{array}{ll} Y = -1121 + 303.4 \ (E) - 8.2 \ (F) + 199.4 \ (H) - 123.5 \\ (E^2) & - 138.2 \ (F^2) & - 9.14 \ (H^2) + 47.3 \ (E^*F) - 10.99 \\ (E^*H) + 18.45 \ (F^*H) & \dots \ (2) \end{array}$ 

The ANOVA of the second-order response surface model evaluated by Fisher's statistical F-test showed that the regression model is highly significant for the production of alkaline protease by B. cereus HP RZ17. The F-value of square regression was higher vis-à-vis the tabulated p < F (Table 3); The results of variables, interaction in different forms that effects of linear and square interaction, the square interaction between among the model terms of E, F, and H variables were significant, while in 2-way interaction the effect of interaction between F and H were found to be non-significant (Fig. 2). By evaluating the coefficient of determination  $(R^2)$  and adjusted  $R^2$  the goodness-of-fit of the model was examined. The higher  $R^2$  represented the maximum variation in observations of the fitted regression equation model.<sup>38,39</sup> The resulting high  $R^{\overline{2}}$  value reflected that the fitted model could explain 95.32% of

Table 3 — AN	OVA c	of a quadrat	tic model f	or alkaline	protease
Source	DE		Adi MS	_KZ17	P-Value
		Auj 55	A00 70	22 <i>c</i> 5	1 - v alue
Model	9	4489.02	498.78	22.65	<0.000
E: Yeast Extract	1	911.30	911.30	41.39	$<\!0.000$
F: Fructose	1	447.45	447.45	20.32	< 0.001
H: pH	1	208.99	208.99	9.49	< 0.012
$E^2$	1	858.33	858.33	38.98	$<\!\!0.000$
$F^2$	1	1074.54	1074.54	48.81	$<\!\!0.000$
$H^2$	1	1203.59	1203.59	54.67	$<\!\!0.000$
EF	1	69.80	69.80	3.17	>0.105
EH	1	60.45	60.45	2.75	>0.129
FH	1	170.11	170.11	7.73	< 0.019
Error	10	220.17	22.02	_	_
Lack-of-Fit	5	219.27	43.85	242.58	$<\!\!0.000$
Pure Error	5	0.90	0.18	—	—
Total	19	4709.19	—	—	—
$R^2 = 95.32\%, R$ (p<0.05 Signification)	<sup>2</sup> (pred. int)	) = 6	4.32%,	$R^2(\mathrm{Adj.})=$	91.12%,



# Surface Plot of Yield vs pH, Fructose

Surface Plot of Yield vs pH, Yeast Extract



Surface Plot of Yield vs Fructose, Yeast Extract



Fig. 2 — The 3D surface plots representing the response of interaction between carbon source - media pH, nitrogen source - media pH and carbon source - nitrogen source on alkaline protease production from *B. cereus* HP\_RZ17

the total variation. The adjusted  $R^2$  corrects the  $R^2$  value for the sample size and the number of terms in the model. The adjusted  $R^2$  value (91.12%) in the present study supported the high significance of the model. The standard error of the regression (*S*) of 4.32 which represents that the observations of model fall in the regression line and show the good relation between predicted and experimental response. These results state that the model provided for the CCD experiment by the response equation are appropriate.<sup>18,39</sup> The 3D-surface plots illustrated the optimum concentrations of each significant variable while interacting with other media variables for the maximum production. The 2D-counter plots of the relationship between fitted responses of two continuous significant variables were observed by contour lines of the points that have the same response value in production. The 3D-surface plots (Fig. 2) and 2D-counter plots (Fig. 3) displayed the various effective regions along with the highest yielding point of the alkaline protease by each variable.

The pH is one of the important physical variables as it controls various enzymatic processes and molecular transportation through the cell membrane. Changes in media pH disturb the equilibrium of the system by fluctuating the H<sup>+</sup> and OH<sup>-</sup> concentration and media nutrients. The optimum pH for most of the alkaliphilic Bacillus strains has been recorded in the range 9.0–11.0.<sup>40</sup> The alkaline protease production is strongly influenced by the various media constituents such as carbon and nitrogen sources. The growth rate and enzyme production rely on the nutritional requirement of strain producing alkaline protease and the nature or type of carbon and nitrogen source may vary as per the requirement of alkaline protease producing organism. The complex medium composed of optimum concentrations of these nutrient sources was found to be the best for the optimum production of alkaline protease.<sup>41</sup> Patel et al.<sup>42</sup> and Solangi et al.<sup>43</sup> reported the optimum alkaline protease production by B. cereus AG1 and B. subtilis in a statistically optimized complex medium containing fructose and yeast extract respectively.

Using the CCD experimental results, Minitab18 (software) predicted the optimum concentrations of yeast extract (E= 0.899%), fructose (F=0.873) and pH (H= 11.25) and their response as alkaline protease yield (133.21 UmL<sup>-1</sup>). The predicted model was analyzed experimentally in triplicate, and the results were as per the predicted yield of alkaline protease (130.72 UmL<sup>-1</sup>) which validates the quadratic model predicted by Minitab at shake flask level.

#### Scale-up at laboratory scale (5L) bioreactor level

Statistically optimized concentrations of the significant variables when used for scale-up from shake flask level to 5L lab bioreactor, a significant increase in the yield of alkaline protease was evident from 130.72 UmL<sup>-1</sup> to 132.84 UmL<sup>-1</sup>. A 1.01 (statistically shake flask to bioreactor) fold or 1.08



Contour Plot of Yield vs pH, Yeast Extract



Contour Plot of Yield vs Fructose, Yeast Extract



Fig. 3 — Counter plots showing the combined effects of carbon source - media pH, nitrogen source - media pH and carbon-nitrogen source on alkaline protease production from *B. cereus* HP\_RZ17

(un-optimized 121.96 UmL<sup>-1</sup> to statistically optimized bioreactor) fold increase in the yield of alkaline protease was obtained due to statistical optimization. Limkar *et al.*<sup>30</sup> reported 1.33 fold increases in the alkaline protease production in *Bacillus* sp. using a statistical optimization approach.

#### Conclusions

The statistical approach facilitates the combinations of experiments to elucidate the significant variables of the production medium for optimum production of alkaline protease in B. cereus HP RZ17. The significant variables initially screened by PBD were further optimized by using RSM. The successfully scale up of statistical-based shake flask process to laboratory scale (5L) bioreactor validated the variables and their concentrations studied at the shake flask level. The values of variables namely veast extract (E), fructose (F) and pH (H) predicted at shake flask level further enhanced the enzyme yield by 1.09 fold (from 121.96 to 132.84 UmL<sup>-1</sup>). This reflects the usefulness of the application of statistical optimization in enzyme production. The experimental results obtained showed a good relationship with the values predicted by Minitab 18 (software). Thus present results illustrated that the statistically optimized bioprocess could be the best suitable approach for enhancing the productivity of alkaline protease at an industrial scale.

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