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Statistical Optimization of Culture Medium for Maximal Growth of Novel Aquaculture Probiotic Strain *Priestia paraflexa* ONG1

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Abstract

This investigation intended to optimize the culture medium for *Priestia paraflexa* ONG1, a fish probiotic, by combining conventional methods with factorial experimental designs. One-variable-at-a-time (OVAT) experiments demonstrated that lactose and tryptone were the preferred carbon and nitrogen sources, respectively, whereas optimal growth occurred at pH 7.6. Plackett-Burman design (PBD) was employed to screen seven variables (lactose, tryptone, yeast extract, NaCl, KH_2PO_4 , MgSO_4 , and MnSO_4) for their influence on probiotic cell density. Statistical analysis revealed yeast extract, MgSO_4 , and MnSO_4 as significant factors ($p < 0.05$), which were subsequently optimized using response surface methodology (RSM). Optimal concentrations for the three factors were identified using a Box-Behnken design (BBD). A predictive model for cell density was then generated using a quadratic equation. The model exhibited a strong fit ($R^2 = 98.73\%$, adjusted $R^2 = 96.45\%$) and identified yeast extract, MnSO_4 , and their quadratic and interaction terms as the most influential factors ($p < 0.05$). The optimized medium, containing lactose (20 g/L), tryptone (2 g/L), yeast extract (8.82 g/L), NaCl (2.5 g/L), KH_2PO_4 (1.25 g/L), MgSO_4 (0.15 g/L), and MnSO_4 (0.5 g/L), resulted in a maximum cell density of 6.5×10^9 CFU/ml, representing a 28.3-fold increase compared to the control nutrient broth medium. This study demonstrated the successful optimization of growth conditions for *P. paraflexa* ONG1, providing valuable insights for the development of this promising aquaculture probiotic.

Keywords: Probiotic, *Priestia paraflexa* ONG1, Plackett-Burman Design, Box-Behnken Design, Medium Optimization, Response Surface Methodology

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INTRODUCTION

Probiotics are beneficial microorganisms that play a crucial role in aquaculture, offering numerous advantages for aquatic animal health and sustainable production.¹ They are widely used as an eco-friendly alternative to antibiotics for controlling fish diseases and improving overall aquaculture performance.^{2,3} Aquaculture commonly employs probiotics from the genera *Bacillus*, *Lactobacillus*, *Lactococcus*, *Enterococcus*, and *Saccharomyces*.⁴ Among these, *Bacillus* species are considered distinctive probiotics in aquaculture due to their ability to form spores, which allows them to survive harsh environmental conditions.⁵ These probiotics possess several advantageous characteristics, including their non-pathogenic nature, lack of toxicity to fish, and the ability to synthesize antimicrobial compounds, which contribute to their superior suitability as compared to other probiotic options in aquaculture.⁶ Environmental conditions, including temperature, pH and stirring intensity, have a substantial impact on the development of probiotic.⁷ Moreover, the composition of the substrate and the availability of nutrients are essential factors in the production of probiotic biomass.⁸ By optimizing probiotic growth, we can produce more viable probiotic cells, which is essential for developing high-quality probiotic products.⁹

Optimal microbial biomass production depends on the appropriate design of the growth medium.¹⁰ To maximize the production of our target products, we experimented with different medium formulations. A traditional one-variable-at-a-time (OVAT) method requires altering a single variable while keeping the others constant. Though simple, it is inefficient due to its high experimental demand and neglect of factor interactions.¹¹ In recent times, different statistical approaches allow researchers to efficiently screen and optimize multiple medium components simultaneously.¹² Accordingly, Plackett-Burman design (PBD) is a statistical tool that employs a fractional factorial design to systematically identify key factors influencing probiotic growth in fermentation media, significantly reducing experimental effort and cost.¹³ This design minimizes bias in linear effect estimation, simplified medium composition, and

facilitates subsequent optimization procedures.¹⁴ While the PBD effectively screened the initial variables, it also set the stage for more detailed optimization techniques, potentially including response surface methodology (RSM) for fine-tuning the identified significant factors.¹⁵ RSM is a well-established modeling technique that is frequently employed to enhance the production of a diverse array of biotechnological products, including biopesticides,¹⁶ exopolysaccharides,¹⁷ antibiotics,¹⁸ enzymes,¹⁹ and biofuels.²⁰

Priestia paraflexa ONG1, a spore-forming probiotic bacterium obtained from the digestive tract of wild Nile tilapia (*Oreochromis niloticus*) exhibits potential probiotic benefits. Importantly, tolerance to acid and bile salt, extracellular enzyme production, adhesion ability, and biofilm formation are essential traits for its potential application as a feed additive in aquaculture. Cultivating this strain at high cell densities is helpful in improving the health and disease resistance of aquatic animals. This study sought to identify the optimal culture conditions that would maximize the cell density of the probiotic *P. paraflexa* ONG1 strain. OVAT was initially employed to identify the optimal pH, carbon, and nitrogen sources as the most influential medium components. Subsequently, a key component within the medium was further investigated using PBD. Lastly, a Box-Behnken design (BBD) was employed to meticulously investigate and fine-tune the medium composition for maximal growth of *P. paraflexa* ONG1.

MATERIALS AND METHODS

Bacterium and seed culture preparation

This study used a gut-associated probiotic strain, *Priestia paraflexa* ONG1 (PQ057050.1), sourced from the intestines of wild Nile Tilapia. The strain was activated twice in nutrient broth at 30 °C before each experiment. The seed medium, consisting of peptone (1%), NaCl (0.5%), and yeast extract (0.2%) was prepared and adjusted pH to 7.0 ± 0.2. The seed culture was cultivated in 20 ml of sterile seed medium within an orbital flask shaker operating at 150 rpm for an incubation period of 16-18 hour at a temperature of 30 °C. A 1% (v/v) inoculum of the seed culture was used to initiate growth in the experimental media for all experiments. Concurrently, the serial dilution and

spread plate method were employed to determine the colony-forming units (CFU) of *P. paraflexa* ONG1 strain in unoptimized media.²¹

OVAT-based media optimization

The impact of individual factors, namely carbon sources, nitrogen sources, and pH, on bacterial cell mass was investigated using the OVAT approach. A preliminary assessment of the carbohydrate utilization capacity of the culture was conducted before implementing the OVAT design. Glucose, sucrose, lactose, cellulose, and starch were chosen as the carbon sources. The basal medium, prepared in 10 ml aliquots, is composed of 0.5% yeast extract, 1% carbon source, and bromocresol purple as a pH indicator.¹¹ A modified cultivation medium, based on the protocol of Vljakov *et al.*,¹⁶ was employed to determine the most suitable carbon source. The following components were included in the medium at the indicated concentrations: carbon source (1%), yeast extract (0.5%), NaCl (0.5%), KH_2PO_4 (0.25%), MgSO_4 (0.03%), and MnSO_4 (0.027%). Glucose, sucrose, lactose, and starch were chosen as carbon sources according to metabolic profile of the culture. A control medium formulated without an added carbon source, while incorporating 0.5% yeast extract, may nonetheless support the proliferation of certain bacterial species due to the presence of residual carbohydrate content in the yeast extract. Medium pH was fixed to 7.0 ± 0.2 before autoclaving the medium. In the subsequent phase of the experiment, lactose was used as carbon source. This study investigated a range of nitrogen sources, including peptone, tryptone, urea, and ammonium sulfate. The remaining

medium components and pH were kept constant. Culture cultivation was carried out using an orbital shaker (150 rpm) at 30 °C for 24 h incubation period. To optimize the pH for cell growth, the media pH was systematically adjusted within the range of 6.0 to 8.0, in increments of 0.4 (6.0, 6.4, 6.8, 7.2, 7.6, and 8.0), using 1 M NaOH and 1 M HCl. Cell density for each experimental condition was determined spectrophotometrically at 600 nm after 24 hour of cultivation.

Plackett-Burman Design (PBD)

The growth of *P. paraflexa* ONG1 was investigated using PBD to determine the significant influencing factors. These factors included: lactose (A), tryptone (B), yeast extract (C), NaCl (D), KH_2PO_4 (E), MgSO_4 (F), and MnSO_4 (G). Two levels were assigned to each variable: a low (-1) and a high (+1) level (Table 1). To evaluate the impact of these factors on the cell density, 12 experimental runs were conducted. Each run the components were formulated using the specific concentrations (Table 2), and the fixed optimum pH was 7.6. A 1% inoculum of *P. paraflexa* ONG1 was added to each sterile media flask, incubated at 30 °C for 24 hour on an orbital shaker, and cell density was determined spectrophotometrically at 600 nm. Statistical analysis was performed to identify significant factors influencing cell growth (Table 3). Factors were considered statistically significant at

Table 1. High and Low level of factors (in g/L) used in PBD

Factor	Coded factor	Low level (-1)	High level (+1)
Lactose	A	2	20
Tryptone	B	2	20
Yeast extract	C	1	10
NaCl	D	2.5	25
KH_2PO_4	E	1.25	5
MgSO_4	F	0.15	0.6
MnSO_4	G	0.13	0.5

Table 2. PBD: Experimental matrix and response values

Run	Factors							Cell density (A_{600})
	A	B	C	D	E	F	G	
1	1	1	-1	1	-1	-1	1	0.52 ± 0.07
2	1	1	1	-1	1	-1	-1	0.53 ± 0.03
3	-1	1	1	1	-1	1	-1	0.53 ± 0.01
4	-1	-1	1	1	1	-1	1	0.57 ± 0.00
5	1	-1	-1	1	1	1	-1	0.52 ± 0.03
6	-1	1	-1	-1	1	1	1	0.57 ± 0.02
7	1	-1	1	-1	-1	1	1	0.63 ± 0.01
8	-1	1	1	-1	1	-1	-1	0.54 ± 0.02
9	-1	1	-1	1	-1	1	1	0.56 ± 0.04
10	1	-1	1	-1	-1	-1	1	0.59 ± 0.02
11	1	-1	-1	1	1	1	-1	0.50 ± 0.05
12	-1	-1	-1	-1	-1	-1	-1	0.42 ± 0.04

A: Lactose; B: Tryptone; C: Yeast extract; D: NaCl; E: KH_2PO_4 ; F: MgSO_4 ; G: MnSO_4

$p < 0.05$. Based on the significance of each factor, high or low levels were selected for subsequent optimization studies. Non-significant factors were excluded from further optimization.¹⁴

RSM: Box-Behnken design (BBD)

A BBD within the RSM framework was employed to optimize the growth of *P. paraflexa* ONG1 by systematically investigating the combined effects of selected variables and determining the optimal conditions.²² The investigation of the three most critical variables involved a design matrix encompassing 15 trials. Each variable under scrutiny was assessed at three distinct levels: low (-1), medium (0), and high (+1) values (Table 4). Cell density after a 24-hour incubation period served as the response variable. A quadratic polynomial regression model was applied to predict the response based on the relationships with the correlated independent variables.²³ This model was subsequently used to determine the optimal conditions. The accuracy of the predicted model was assessed through independent triplicate experiments conducted under optimized cultivation conditions. Cultures were grown in optimized medium, and cell viability was validated using the viable count method.

Statistical analysis

Statistical analyses were performed using SPSS software (IBM, USA). All experiments were conducted in triplicate, and data are presented as mean \pm standard deviation. One-way ANOVA

followed by Duncan's Multiple Range Test (DMRT) was used to determine statistically significant differences. PBD and RSM designs were generated and analyzed using Minitab 22 (Minitab Inc., USA).

RESULTS AND DISCUSSIONS

Optimizing Carbon, Nitrogen sources, and pH for *P. paraflexa* ONG1 cell density production

Bacterial growth is intricately linked to the availability and utilization of both carbon and nitrogen sources. Diverse carbon and nitrogen substrates exhibit varying degrees of suitability for biomass production.^{24,25} Various carbon and nitrogen sources were evaluated to determine the most significant factors for maximizing high cell density production in *P. paraflexa* ONG1. Carbohydrate utilization tests indicated that probiotic bacteria could ferment glucose, sucrose, lactose, and starch. As a result, the selected carbon sources were the sole focus of the subsequent optimization. Lactose yielded the highest cell density ($OD_{600} = 0.40 \pm 0.005$) among the four carbon sources tested. Starch, sucrose, and glucose resulted in lower densities of 0.35 ± 0.028 , 0.32 ± 0.003 , and 0.25 ± 0.011 , respectively. Statistical analysis confirmed that lactose was the preferred carbon source for *P. paraflexa* ONG1 ($p < 0.05$). The observed preference for lactose aligns with the findings of Niekerk and Pott,²⁶ who demonstrated the superior biomass production of *Bacillus circulans* DSM-11 on lactose compared with glucose, achieving a 1.7-fold increase in biomass

Table 3. Statistical analysis of factors influencing high cell density production of *P. paraflexa* ONG1 using PBD data

Variables	Effect	Coefficient	t-value	p-value	Significance
Model		0.54000	112.67	0.000	**
A	0.02000	0.01000	1.97	0.121	
B	0.01000	0.00500	0.98	0.381	
C	0.06300	0.03150	6.00	0.004	**
D	-0.00800	-0.00400	-0.76	0.489	
E	0.02125	0.01062	2.09	0.105	
F	0.04700	0.02350	4.48	0.011	*
G	0.07375	0.03687	7.25	0.002	**
R ²	0.9630				
Adj-R ²	0.8983				

A: Lactose; B: Tryptone; C: Yeast extract; D: NaCl; E: KH_2PO_4 ; F: $MgSO_4$; G: $MnSO_4$

* represents $p < 0.05$; ** represents $p < 0.01$

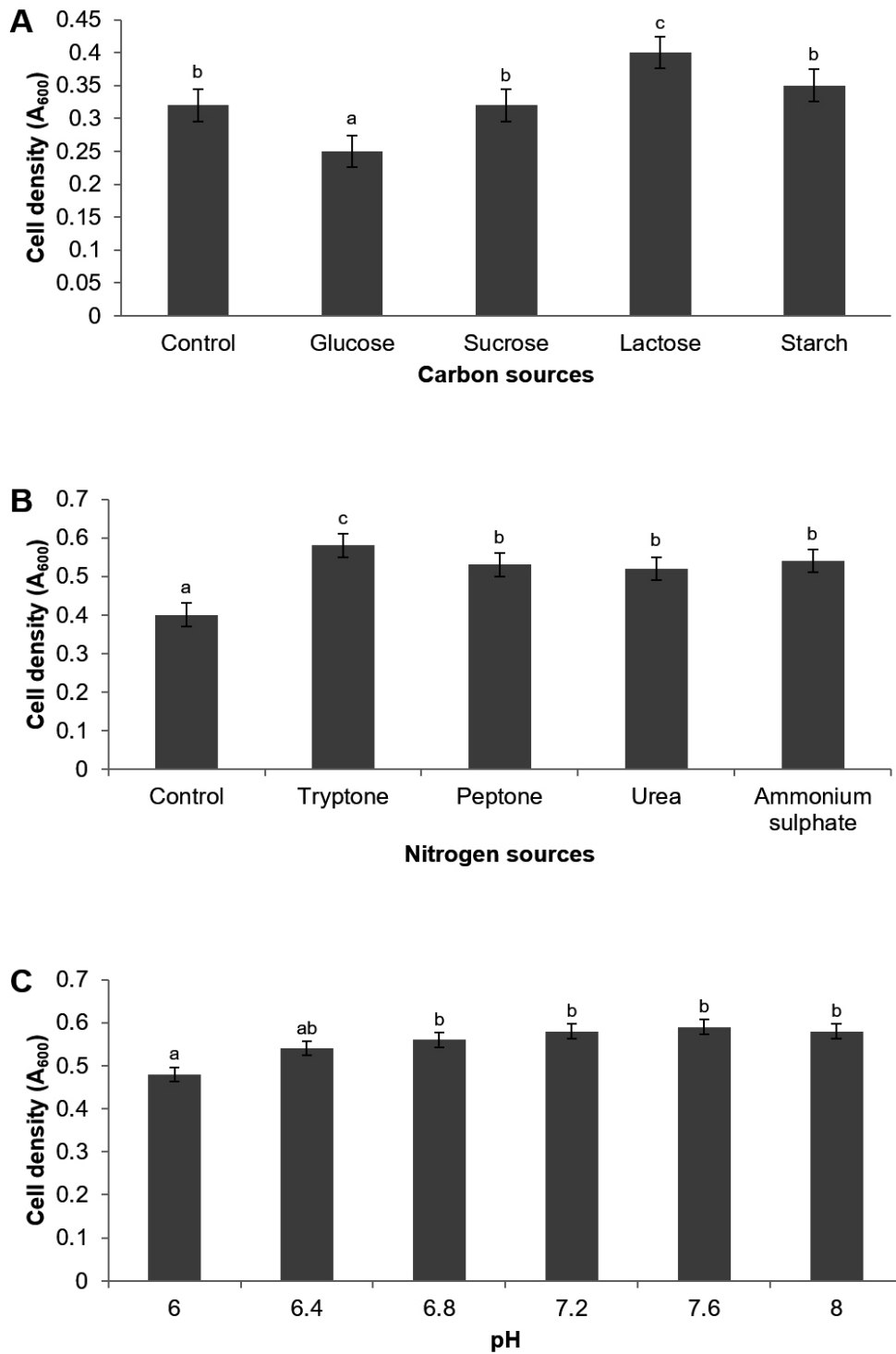


Figure 1. Impact of Carbon (A) Nitrogen (B) and pH (C) on high cell density cultivation of *P. paraflexa* ONG1. Data are reported as mean \pm SE based on three independent experiments. Significant differences ($p < 0.05$) are indicated by different letters (a, b, and c)

yield. Interestingly, lactose supplementation in a growth medium for *Priestia flexa* N7 significantly enhanced hyaluronic acid (HA) production, with concentrations reaching 601.6 mg/L in a recent study.²⁷ Figure 1A and 1B illustrate how various carbon and nitrogen sources influence cell density.

Tryptone was the most effective nitrogen source, leading to the highest growth (0.58 ± 0.017) at 600 nm, followed by ammonium sulfate (0.54 ± 0.023), peptone (0.53 ± 0.011), and urea (0.52 ± 0.040). Peptone and yeast extract are the favored nitrogen sources for *Priestia* species. Remarkably, *P. filamentosa* and *P. aryabhatai* demonstrated superior growth capabilities when provided with organic nitrogen sources compared to inorganic nitrogen sources.²⁸ Our findings concur with this study, likely due to the use of tryptone as an organic nitrogen source. Supplementation of growth medium with urea led to a slight reduction in biomass production by *P. paraflexa* ONG1. Conversely, 0.1% urea emerged as the optimal nitrogen source for maximizing curdian production by *Priestia megaterium*, whereas the strain demonstrated a limited preference for

peptone.²⁹ Probiotic strain *P. paraflexa* ONG1 exhibits the metabolic versatility to utilize a variety of nitrogen sources, including both organic and inorganic compounds. This adaptability enables them to successfully colonize and persist in environments with limited nutrient availability.

Optimizing probiotic growth requires careful consideration of not only the specific media composition but also physicochemical properties, such as pH. The optimal cell growth (0.59 ± 0.00) occurred at pH 7.6, with decreasing values observed on either side of this point (Figure 1C). Minimal growth (0.48 ± 0.02) was recorded at pH 6.0. Notably, the organism

Table 4. Experimental design parameters for RSM experiments

Variables	Units	Symbol code	Level		
			+1	0	-1
Yeast extract	g/L	X_1	10	5.5	1
MgSO ₄	g/L	X_2	0.6	0.375	0.15
MnSO ₄	g/L	X_3	0.5	0.315	0.13

Table 5. Experimental design matrix for BBD

Run	Factors			Cell density (A_{600})
	X_1	X_2	X_3	
1	0	-1	-1	1.02 ± 0.02
2	-1	-1	0	0.96 ± 0.01
3	0	0	0	1.12 ± 0.04
4	1	-1	0	1.14 ± 0.04
5	0	1	-1	1.08 ± 0.02
6	1	0	-1	1.04 ± 0.07
7	-1	1	0	0.98 ± 0.04
8	0	0	0	1.12 ± 0.02
9	-1	0	-1	0.92 ± 0.05
10	0	-1	1	1.18 ± 0.02
11	-1	0	1	0.98 ± 0.01
12	0	1	1	1.19 ± 0.05
13	0	0	0	1.11 ± 0.03
14	1	1	0	1.11 ± 0.04
15	1	0	1	1.19 ± 0.01

X_1 : Yeast extract; X_2 : MgSO₄; X_3 : MnSO₄

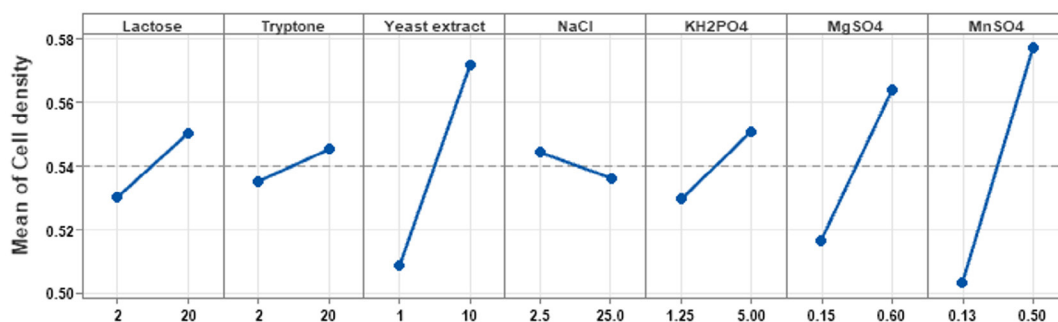


Figure 2. Visualizing the main effects of each variable on *P. paraflexa* ONG1 cell density

exhibited minimal variation in cell density between pH 6.4 and 8.0 suggesting that the probiotic can thrive across a broad pH range. Our results align with a previous study on *P. megaterium* gasm, an elastase-producing bacterium from processed meat, which found that pH 8 is optimal for elastase production by this strain.³⁰ Concurrently, prior research corroborates the observation that *Priestia* species exhibit a remarkable capacity to thrive within a diverse pH range, encompassing slightly acidic to alkaline environments.³¹⁻³³

Plackett-Burman design for identifying key factors

PBD was implemented to determine the influence of each individual medium component on achieving high cell densities of *P. paraflexa* ONG1. Seven variables (lactose, tryptone, yeast extract, NaCl, KH_2PO_4 , MgSO_4 , and MnSO_4) were examined at two levels (-1, 1) (Table 1), leading to 12 experimental runs. Three replicates were conducted for each experimental condition to improve the reliability of the data. The resulting growth yields, measured as optical density at 600 nm, varied from 0.42 to 0.63 (Table 2). Table 3 presents the effects of the seven variables. The data revealed the following effects for each component: Lactose 0.020, tryptone 0.010, yeast extract 0.063, NaCl -0.008, KH_2PO_4 0.021, MgSO_4 0.047, and MnSO_4 0.073. Among the seven variables examined, NaCl exhibited a negative correlation with probiotic growth (Figure 2). Elevating the NaCl

concentrations within the tested range of 2.5 to 25 g/L resulted in a decline in cell density production. The standardized impact of each variable is visually represented in the Pareto chart (Figure 3). The vertical line on the chart indicates the statistical significance. Variables with columns extending to the right of this line have a substantial influence on bacterial growth. Statistical analysis revealed that yeast extract, MgSO_4 , and MnSO_4 significantly influenced high cell density production ($p < 0.05$). Additionally, as indicated by the R^2 value of 0.963, the model effectively accounts for 96.3% of the response variability. The high adjusted R^2 value of 0.898 (indicating model significance) supports the suitability of the PBD. Thus yeast extract, MgSO_4 , and MnSO_4 were chosen for subsequent RSM analysis. Shi *et al.*³⁴ highlighted the significant impact of magnesium sulfate on the colony formation of *Bacillus velezensis* BHZ-29, alongside other contributing factors. Another investigation, employing the Plackett-Burman experimental design, demonstrated that yeast extract was a significant factor in promoting spore production of *B. coagulans* X26.³⁵ Manganese, supplied as MnSO_4 , is an essential micronutrient for *Bacillus* growth and serves as a key inducer of sporulation. Prior investigations have indicated the potential for manganese supplementation to augment spore formation efficiency,³⁶ enzyme production³⁷ and bacterial biomass production.³⁸

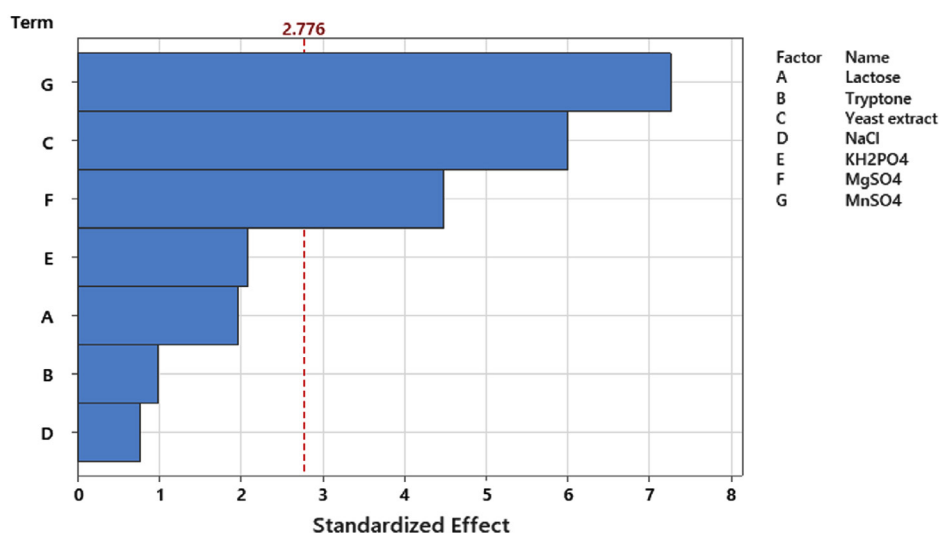


Figure 3. Visualization of standardized effects on *P. paraflexa* ONG1 cell density production using a Pareto chart

Table 6. Analysis of variance for Box-Behnken design (BBD) experiments in the optimization of *P. paraflexa* ONG1 cell density

Sources	DF	Adj SS	Adj MS	F-value	p-value	Significance
Model	9	0.10639	0.01182	43.25	0.000	**
Linear	3	0.08045	0.02682	98.11	0.000	**
X_1	1	0.05120	0.05120	187.32	0.000	**
X_2	1	0.00045	0.00045	1.65	0.256	
X_3	1	0.02880	0.02880	105.37	0.000	**
Square	3	0.02267	0.00756	27.64	0.002	**
X_1^2	1	0.02194	0.02194	80.27	0.000	**
X_2^2	1	0.00023	0.00023	0.85	0.400	
X_3^2	1	0.00019	0.00019	0.68	0.448	
2-Way interaction	3	0.00328	0.00109	3.99	0.085	
X_1X_2	1	0.00063	0.00063	2.29	0.191	
X_1X_3	1	0.00203	0.00203	7.41	0.042	*
X_2X_3	1	0.00063	0.00063	2.29	0.191	
Error	5	0.00137	0.00027			
Lack-of-Fit	3	0.00130	0.00043	13.00	0.072	
Pure error	2	0.00007	0.00003			
Total	14	0.10776				
R^2		0.9873				
Adj- R^2		0.9645				

X_1 : Yeast extract; X_2 : $MgSO_4$; X_3 : $MnSO_4$; DF: Degree of freedom; SS: Sum of squares; MS: Mean square; *represents $p < 0.05$; **represents $p < 0.01$

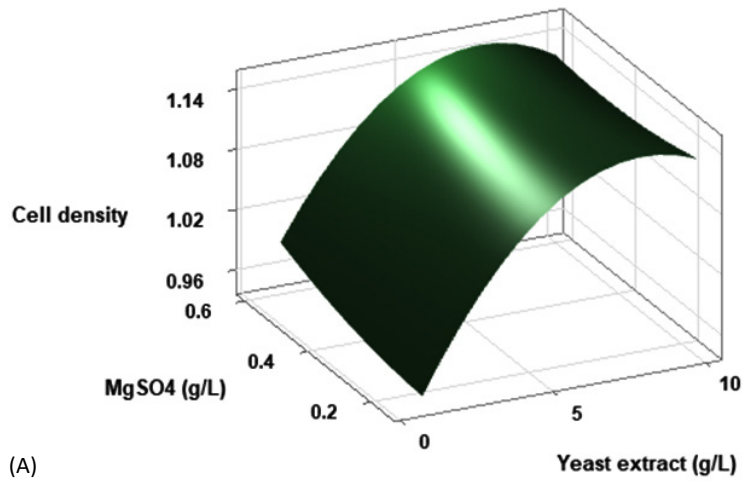
Response surface analysis

The approximate optimal region was identified using a response surface design. BBD was specifically utilized to discover the optimal concentrations of yeast extract (1-10 g/L), $MgSO_4$ (0.15-0.6 g/L), and $MnSO_4$ (0.13-0.5 g/L). The coded levels and their corresponding concentrations for the three variables used in the experimental investigation are summarized in Table 4. Table 5 presents the experimental design matrix, comprising 15 runs, including three center point replicates, and their corresponding responses. A maximum cell density 1.19 OD_{600} was observed in run 15, corresponding to a combination of 10 g/L yeast extract, 0.375 g/L $MgSO_4$, and 0.5 g/L $MnSO_4$. Multiple regression analysis was used to fit a quadratic polynomial model to the experimental data for predicting *P. paraflexa* ONG1 cell density. The fitted equation is:

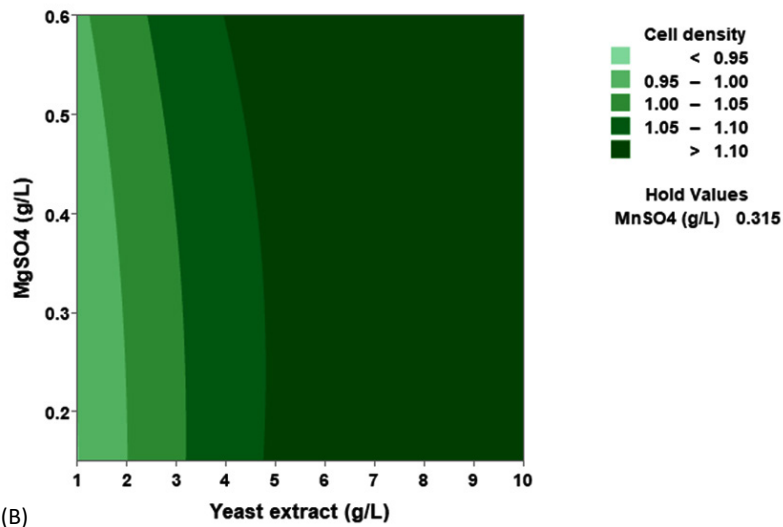
$$Y = 0.7764 + 0.05577 X_1 + 0.079 X_2 + 0.419 X_3 - 0.003807 X_1^2 + 0.156 X_2^2 - 0.207 X_3^2 - 0.01235 X_1 X_2 + 0.02703 X_1 X_3 - 0.300 X_2 X_3$$

In the equation, Y denotes the cell density (OD_{600}), while X_1 , X_2 , and X_3 correspond to yeast extract, $MgSO_4$, and $MnSO_4$, respectively.

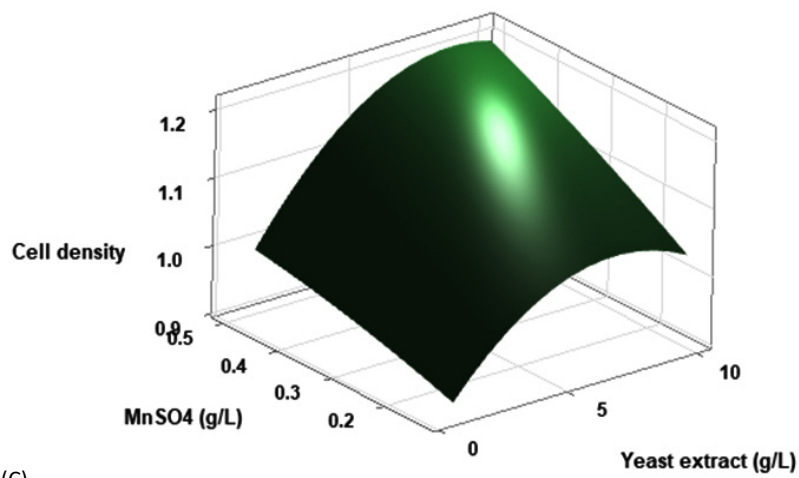
Statistical significance was examined through the application of Analysis of variance (ANOVA), with subsequent utilization of an F-test. The calculated F-value of 43.25, as presented in Table 6, indicates a highly significant model (p-value 0.000). The developed regression model, exhibiting an R^2 of 98.73% and an adjusted R^2 of 96.45%, effectively predicted the cell density of *P. paraflexa* ONG1, accounting for over 95% of the variability in the data. Furthermore, a statistically non-significant lack-of-fit (p-value 0.072) indicated that the model provided a good fit to the experimental data. The magnitude of the p-value is inversely related to the term's significance, with lower values indicating stronger evidence against the null hypothesis and potentially elucidating variable interactions.¹⁴ Analysis of Table 6 revealed that X_1 , X_3 , and X_1^2 were the most influential factors in this model, each



(A)



(B)



(C)

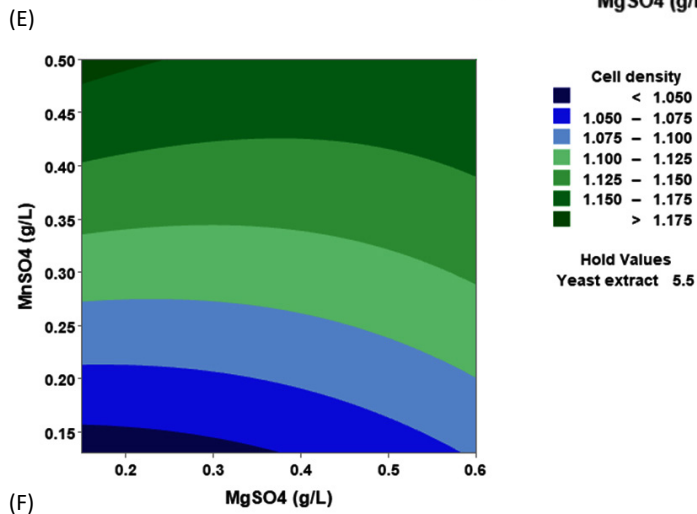
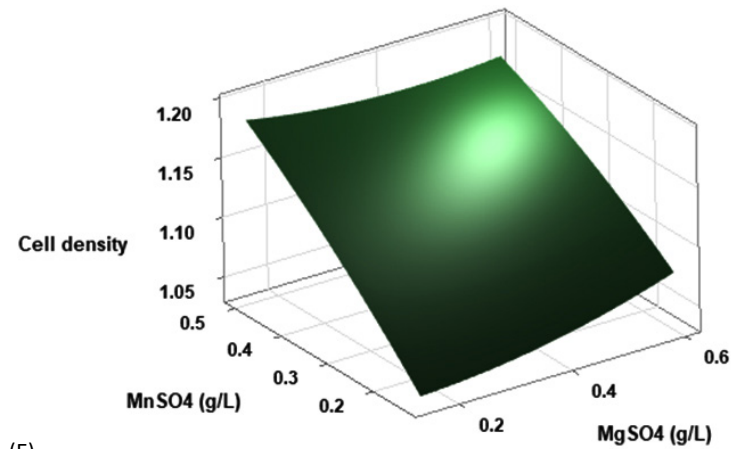
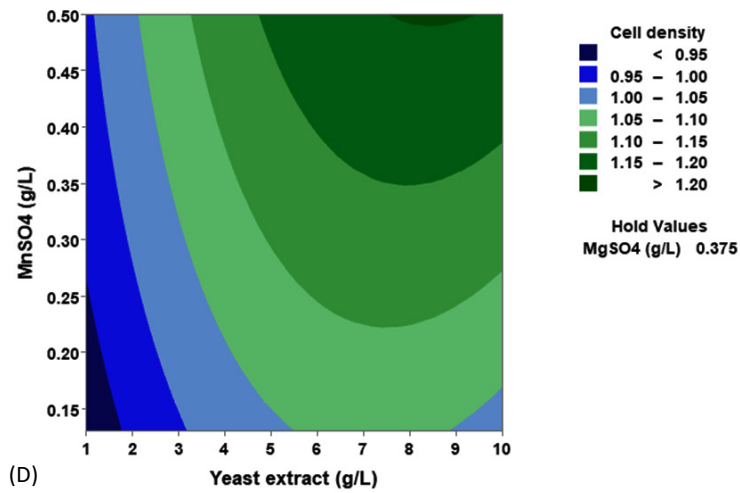


Figure 4. Three-dimensional response surface and two-dimensional contour plots depicting the interactive effects of yeast extract (X_1), $MgSO_4$ (X_2), and $MnSO_4$ (X_3) on the cell density of *P. paraflexa* ONG1. (A, B) Interaction between X_1 and X_2 with X_3 held constant at its central value. (C, D) Interaction between X_1 and X_3 with X_2 maintained at its central value. (E, F) Interaction between X_2 and X_3 with X_1 fixed at its central value

demonstrating a statistically significant association ($p < 0.01$). The insignificance of the interaction terms is further supported by the elevated p-value ($p > 0.05$), with the exception of the interaction between X_1 and X_3 (p-value 0.042). To gain a visual understanding of the interactive effects between the two variables and to identify the optimal concentration combinations for maximizing cell density, 3D response surface and contour plots were constructed using Minitab statistical software (Figure 4). Graphical representations based on the model equation were generated, which displayed convex response surfaces. Each graph illustrates the interactive effects of the two independent variables, whereas the remaining variable is fixed at a specific level. Cell density exhibited a positive correlation with yeast extract concentration within the intervals of 4.8 g/L to 10 g/L and 7.5 g/L to 8.82 g/L, as shown in Figures 4A, 4B, 4C and 4D. However, a further increase in yeast extract concentration beyond this point resulted in a decrease in cell density. The concentration-dependent response was investigated for $MgSO_4$ and $MnSO_4$, while maintaining the yeast extract concentration at the central level. Likewise, $MgSO_4$ demonstrated a similar pattern, with maximum cell density production at 0.20 g/L, followed by a reduction. Concurrently, varying concentrations of $MgSO_4$, as shown in Figure 4A and 4B, did not yield a statistically significant enhancement in cell density. The data indicates that elevated levels of the independent variables (yeast extract and $MgSO_4$) negatively impacted cell density. In contrast, for $MnSO_4$, maximum growth was observed at a higher concentration (0.5 g/L) (Figure 4E and 4F). The model predicted that a combination of yeast extract (8.82 g/L), $MgSO_4$ (0.15 g/L), and $MnSO_4$ (0.5 g/L) would maximize probiotic growth, resulting in a predicted cell density of 1.22 Optical Density at 600 nm.

Following cultivation in the optimized medium and under the optimized conditions, *P. paraflexa* ONG1 exhibited a maximum cell density of 6.5×10^9 CFU/ml. This represents a 28.3-fold enhancement compared to the cell density obtained in nutrient broth medium, which was 2.3×10^8 CFU/ml. The significant increase in cell density observed in the optimized medium strongly suggests that it provides an environment conducive to the growth and

metabolism of *P. paraflexa* ONG1. This indicates that the optimized medium likely contained a well-balanced combination of nutrients that effectively meet the specific requirements of this organism. Using the Box-Behnken experimental approach, Eyahmalay *et al.*³⁹ investigated a simple growth medium formulated with lactose (76 g/L), soybean meal (72 g/L), yeast extract (2 g/L), and $MgSO_4$ (0.7 g/L). This optimization process resulted in a significant increase in *Lactobacillus casei* cell biomass, reaching a 164.6% increase. Shi *et al.*³⁴ utilized PBD and BBD to optimize the growth medium for *Bacillus velezensis* BHZ-29, resulting in an increased viable cell count from 7.83×10^9 to 3.39×10^{10} CFU/ml. Information on the optimal growth conditions for *Priestia* species is scarce. To the extent of our current understanding, this investigation signifies the first attempt to optimize the growth of *P. paraflexa*, a promising aquaculture probiotic, utilizing Response Surface Methodology (RSM).

CONCLUSION

This research effectively employed statistical techniques, particularly RSM, to enhance *P. paraflexa* biomass production by optimizing growth conditions and nutrient composition. PBD identified yeast extract, $MgSO_4$, and $MnSO_4$ as significant factors that influence cell density. The subsequent BBD and fitted quadratic polynomial models further refined the optimum levels of these variables. Based on the BBD experimental matrix analysis, the model demonstrated significant effects for the linear terms X_1 and X_3 , the quadratic term X_1^2 , and the interaction term X_1X_3 . Other interaction terms and quadratic terms were not statistically significant. According to the model predictions, the ideal concentrations were determined to be 8.82 g/L for yeast extract, 0.15 g/L for $MgSO_4$, and 0.5 g/L for $MnSO_4$. These optimal levels were projected to yield a maximal cell density of 1.22 optical density units. Validation experiments conducted in triplicate confirmed this prediction, yielding an average cell density of 1.19 optical density units and a viable count of 6.5×10^9 CFU/ml, demonstrating the precision and dependability of the model. This research provides crucial findings on the ideal cultivation conditions for *P. paraflexa* ONG1, significantly enhancing our

limited understanding of *Priestia* species. The significant enhancement in cell density achieved through this optimization process demonstrates the potential for improved production of this promising aquaculture probiotic. These findings could contribute to the development of more efficient and cost-effective probiotic production methods, benefiting the aquaculture industry.

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CONFLICT OF INTEREST

The authors declare that there is no conflict of interest.

AUTHORS' CONTRIBUTION

Both authors listed have made substantial, direct, and intellectual contributions to the work and approved it for publication.

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DATA AVAILABILITY

All datasets generated or analyzed during this study are included in the manuscript.

ETHICS STATEMENT

Not applicable.

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