



Enhancement of BLIS production by *Pediococcus acidilactici* kp10 in optimized fermentation conditions using artificial neural network

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1 **Enhancement of BLIS production by *Pediococcus acidilactici* kp10 in**
2 **optimized fermentation conditions using artificial neural network**

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34 **Running title:** BLIS Production by *Pediococcus acidilactici* kp10

35 **Abstract**

36 The present work was aimed at enhancing the production of BLIS produced by
37 *Pediococcus acidilactici* Kp10 through optimization of fermentation parameters. M17
38 was chosen in preliminary study as a culture medium since the production of BLIS
39 was nine times higher (1,427.7 AU/mL) compared to that produced by MRS (160
40 AU/mL). The fermentation parameters such as temperature, inoculum size, buffer
41 strength, concentration of tween 80 and agitation speed were screened using 2 level
42 half-factorial design. BLIS production influenced by three most significant factors
43 which identified as temperature, inoculum size and agitation speed were further
44 optimized using artificial neural network (ANN). ANN predicted maximum activity
45 of 5,262.64 AU/mL was obtained at optimum conditions of 120 rpm, 3% and 28.5 °C.
46 The observed BLIS activity at the predicted optimum levels of the tested variables in
47 ANN was 5,118.5 AU/mL which was close to the predicted BLIS activity. Increased
48 BLIS activity in the final solution resulted from the optimized process would reduce
49 the downstream step such as concentrating purified product during purification.

50

51 **Keywords:** Bacteriocin-like inhibitory substance, optimization, artificial neural
52 network, natural preservative, antibacterial spectrum, biological origin

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57 **Introduction**

58 Current consumer demands for commercial natural preservative processed foods has
59 created considerable interest in the search for safe and food-grade preservatives of
60 biological origin. Antibacterial peptides or bacteriocins produced by many strains of
61 lactic acid bacteria (LAB) have been used as food preservatives for many years
62 without any known adverse effects. Bacteriocins, especially those with broad
63 antibacterial spectrum are bactericidal against food spoilage and many pathogenic
64 bacteria.

65

66 Specific requirements with reference to the production of bacteriocins have
67 been reported ¹⁻³. Bacteriocin titres can be modified by altering the cultivation
68 conditions of the producing bacterium and certain combinations of influencing factors
69 need to be optimized to enhance bacteriocin production ⁴. With regards to the
70 complexity of the factors within the food environments itself an in-depth knowledge
71 of the interactions of these factors influencing the production of bacteriocin need to
72 be understood for subsequent application in the optimization process. Most studies
73 carried out to date claimed validation by statistical analysis and the combination of
74 variables, their values and limits were arbitrarily chosen based primarily on personal
75 experience ⁵.

76

77 Conventional methods in fermentation optimization require treating each
78 factor separately which are laborious, incomplete and time consuming. If several
79 factors are to be considered simultaneously their interactions are not discernible even
80 for the dominant ones. These conventional approaches did not yield reliable results
81 either. In this respect experimental factorial design have been successfully applied for

82 optimization of various biomanufacturing processes ⁶, which could also be used to
83 investigate the interacting factors ^{7,8}.

84

85 Artificial neural network (ANN) has recently emerged as one of the most
86 efficient methods for empirical modelling and prediction in solving complex systems
87 such as bacteriocin production ^{7, 9}. Several studies have demonstrated that the
88 accuracy of prediction of ANN models were far more superior compared to RSM
89 using the same experimental design ^{10,11}. ANN does not require prior specification of
90 suitable fitting function. ANN has universal approximation capability which means it
91 can approximate almost all kinds of non-linear functions including quadratic
92 functions ¹². The ability of ANN to predict the process characteristics with little prior
93 knowledge is desirable which simplify their implementation and increases their
94 modeling potential. This property makes ANN a powerful and flexible tool that is
95 well-suited to modeling biochemical processes ¹³.

96

97 Reports on multiple factors affecting the production of bacteriocins are
98 relatively scarce and the optimization of bioprocess parameters for enhanced
99 production of bacteriocin remain elusive ¹⁴. Improvement of bacteriocin production
100 would therefore require a comprehensive understanding of the factors affecting their
101 production ¹⁵. Thus the objective of this study was to optimize the fermentation
102 parameters for enhancement of production of BLIS by *P. acidilactici* Kp10. This
103 strain has demonstrated a high antimicrobial activity against *Listeria monocytogenes*,
104 a virulent food-borne pathogen ¹⁶. The effects of fermentation parameters such as
105 temperature, inoculum size, buffer strength, tween 80 concentration and agitation
106 speed on BLIS production were evaluated using the half factorial design. The most

107 significant factors were then optimized using the ANN to predict the enhanced
108 bacteriocin production by *P. acidilactici* Kp10.

109

110 **Materials and methods**

111 **Bacterial strain, medium selection and inoculum preparation**

112 *P. acidilactici* Kp10, isolated from dried curd was used in this study ¹⁶. Two types of
113 medium, MRS and M17 were initially tested in this study. MRS broth (Merck,
114 Darmstadt, Germany) consisted of peptone (10 g/L), meat extract (8 g/L), yeast
115 extract (4 g/L), glucose (20 g/L), di-potassium hydrogen phosphate (2 g/L), Tween 80
116 (1 g/L), di-ammonium hydrogen citrate (2 g/L), sodium acetate (5 g/L), magnesium
117 sulphate (0.2 g/L) and manganese sulphate (0.04 g/L). M17 broth (Merck, Darmstadt,
118 Germany) consisted of peptone (10 g/L), yeast extract (2.5 g/L), meat extract (5 g/L),
119 lactose (5 g/L), ascorbic acid (0.5 g/L), sodium β -glycerophosphate (19 g/L) and
120 magnesium sulphate (0.25 g/L). Three different temperatures (30 °C, 35 °C and 37
121 °C) were employed to select the preferred complex medium for BLIS production by *P.*
122 *acidilactici* Kp10.

123 Primary cultivation was prepared by taking a single bacterial colony from the
124 agar plate and grown in 10 mL tubes each containing M17 broth and incubated at 37
125 °C for 24 h. This initial culture (1 % v/v) was then inoculated into a 50 mL Falcon TM
126 tube containing 10 mL of M17 medium and incubated at 37 °C for 24 h. This culture
127 was used as a standard inoculum throughout the study.

128

129 **Determination of antimicrobial spectrum**

130 Antimicrobial activity of the isolates was determined by the agar well diffusion
131 method ¹⁷ using cell-free culture supernatants (CFCS). The isolates were grown in

132 M17 broth at 30 °C for 24 h and the cultures were centrifuged at 12,000 ×g for 20
133 min at 4°C (rotor model 1189, Universal 22R centrifuge, Hettich AG, Switzerland).
134 Aliquots of supernatant in two-fold dilution (100 µL) were placed in wells (6 mm
135 diameter) of cooled soft agar plates (25 mL) previously seeded (1% v/v) with the
136 actively growing test strain (*Listeria monocytogenes* ATCC 15313). The plates were
137 incubated at 37 °C for 24 h for the growth of *L. monocytogenes* as the target
138 microorganism. After 24 h, the growth inhibition zones were measured, and
139 antimicrobial activity (AU/mL) was determined as described by Parente, Brienza,
140 Moles, & Ricciardi,¹⁸. It is based on results from 2-fold dilutions of CFCS, which are
141 either as a CDA (with the critical dilution being the last dilution which produced a
142 zone larger than well diameter) or as a quantitative assay. A standard curve is
143 prepared using a series of 2-fold dilutions of BLIS solution. The response (*R*) is
144 calculated either as the diameter or as the area of the inhibition zone corrected for the
145 diameter or area of the well. The dose (*d*) is the amount of BLIS pipetted into each
146 well (100 µL 1/D). The dose/response curve is calculated according to equation 1:

$$147 \quad R = a + b \log (d) \quad (1)$$

148

149 The critical dose (*CD*) is defined as the amount of BLIS solution (in mL)
150 corresponding to a null inhibition zone and calculated by extrapolating the
151 dose/response curve. The titer, in AU/mL, is calculated as the reciprocal of the CD.
152 The standard curve, relating dose in AU to *R*, had a 0 intercept and the same slope of
153 Equation 1. The activity in AU/mL of a sample can be calculated using equation 2.

$$154 \quad \text{AU/mL} = (1000/100) D 10^{(R/b)} \quad (2)$$

155

156

157 Experimental design

158 Two types of media (MRS and M17), normally used for LAB cultivation were
159 initially tested for the growth of *P. acidilactici* Kp10 and BLIS production. The two-
160 level half factorial design was initially used to screen important factors that affect
161 BLIS production by *Pediococcus acidilactici* Kp10. The five factors considered were
162 temperature (20 °C and 37 °C), inoculum size (1% and 10%), buffer strength (sodium
163 β -glycerophosphate concentration, 60 μ M and 180 μ M), Tween 80 concentration (0.16
164 g/L and 1.6 g/L) and agitation speed (0 rpm and 200 rpm). Each factor was tested
165 with an equal number of repetitions at high and low levels. A total of 16 experiments
166 were run in duplicates (Table 1) .

167

168 For further optimization of the culture conditions, a total of 34 experimental
169 runs were designed according to Box-Wilson (BW) 2^3 full factorial central composite
170 design (CCD) with three most significant factors (temperature, inoculum size and
171 agitation speed) selected from the initial screening results (Table 2). In the application
172 of ANN, the experimental data obtained from the two-level half factorial design were
173 analyzed using Intelligent Problem Solver (STATISTICA software version 7) to
174 construct the regression based networks from the data. A total of 175 different trained
175 networks were observed for selection on the basis of the highest coefficient of
176 correlation determination (R^2) and the lowest selection error. A Multilayer Perceptron
177 Network (MPN) was selected from the above. Back Propagation (BP) and Conjugate
178 Gradient Descent (CGD) logarithms were used in the training of neural network on
179 the basis of varying input/output pair data sets. The eight experiments were carried
180 out for the selection group, 18 for the training group and 8 for the testing group
181 (Table 3). The network developed in this study comprised of three layers - an input

182 layer with three neurons, one hidden layer with nine neurons and an output layer with
 183 one neuron. Determination of the optimum topology was based on minimum error of
 184 testing. The ideal network was used for the prediction and optimization of BLIS
 185 production.

186

187 **Optimization capability of ANN**

188 The analysis of variance (ANOVA) was employed to determine the
 189 significance of the model parameters. Adjusted R^2 , absolute average deviation (AAD)
 190 and root mean square error (RMSE) were calculated in addition to the R^2 for
 191 prediction of bacteriocin production by the ANN. R^2 was calculated using equation
 192 (3):

$$R^2 = \frac{\sum_{i=1-n} (X_i - y_{i.exp})^2}{\sum_{i=1-n} (\bar{Y}_i - y_{i.exp})^2}$$

193

194

(Equation 3)

195

196 Where;

197 X_i : BLIS production

198 $y_{i.exp}$: experimental BLIS production and

199 \bar{y}_i : average observed BLIS production

200 The adjusted R^2 was calculated using equation (4):

$$Adjusted R^2 = 1 - [(1 - R^2) \times \frac{N - 1}{N - K - 1}]$$

201

202

203 (Equation 4)

204 Where;

205 N: total number of observations and

206 K: number of input variables

207

208 AAD was calculated using equation (5).

$$AAD = \left\{ \left[\sum_{i=1}^P (|y_{i,exp} - y_{i,cal}| / y_{i,exp}) / P \right] \right\} \times 100$$

209

210 (Equation 5)

211 Where;

212 $y_{i,exp}$ and $y_{i,cal}$: experimental and calculated responses and

213 p: number of experiments

214

215 RMSE was calculated using equation (6).

216

$$RMSE = \sqrt{\frac{\sum (y_{i,exp} - y_{i,cal})^2}{n}}$$

217

218 (Equation 6)

219 Where;

220 $y_{i, exp}$: experimental response

221 $y_{i, cal}$ is the calculated response and p is the number of experiments

222

223 Analytical procedures

224 BLIS production by *P. acidilactici* Kp10 was tested by the agar well diffusion method

225 ¹⁹ using CFCS. CFCS was prepared by growing *P. acidilactici* Kp10 in M17 broth at

226 30 °C for 24 h. Cells produced were subsequently separated by centrifugation at
227 $12,000 \times g$ for 20 min at 4 °C. Each aliquot of supernatant in two-fold dilution (100
228 μL) were placed in wells (6 mm diameter) of cooled soft agar plates (25 mL)
229 previously seeded (1% v/v) with the actively growing *L. monocytogenes* ATCC
230 15313. The plates were then incubated at 37 °C for 24h. After that, the diameters of
231 the growth inhibition zones were measured and the antimicrobial activity, defined as
232 the mean reciprocal of the highest dilution showing inhibition of the indicator lawn,
233 was expressed in Activity Units (AU) per mL (AU/ mL). Cell growth was measured
234 by optical density of the culture at 650_{nm} using UV/VIS spectrophotometer (Perkin
235 Elmer, Lambda 25, USA).

236

237 **Result and discussion**

238 LAB are fastidious microorganisms with respect to their nutrient requirement which
239 require a rich medium containing yeast extract and protein hydrolysates for good
240 growth and bacteriocin production^{16,20}. Therefore, a suitable medium is essentially
241 for a good growth and bacteriocin production. For all temperatures tested, BLIS
242 production by *P. acidilactici* Kp10 in M17 medium was substantially higher
243 compared to that produced in MRS medium (Table 4). For both media, the highest
244 BLIS production was obtained at 30 °C and a drastic reduction of BLIS production
245 was observed at 37 °C. BLIS production using M17 medium was nine times higher
246 (1,427.7 AU/mL) than that obtained using MRS medium (160 AU/mL). Hence, M17
247 medium was selected as a basal medium for the subsequent experiments.

248

249 Results from the screening of culture conditions using two-level half factorial
250 design showed that the highest BLIS production (run 9) was at 20 °C with 10 g/L of

251 tween 80 while the lowest production (run 8) was obtained at 37 °C with 0.16 g/L of
252 tween 80 (Table 1). Other parameters - inoculum size (10%) buffer strength (60 µM)
253 and agitation speed (200 rpm) remained the same for the highest and the lowest
254 production of BLIS. The culture pH which ranged between 6 (experimental run 6) and
255 7.01 (experimental run 11) was well correlated with respect to the buffer strength.
256 Analysis of variance (ANOVA) showed that temperature, agitation speed and
257 inoculum size were the most significant factors influencing BLIS production by *P.*
258 *acidilactici* Kp10 (Table 5). On the other hand, buffer strength and tween 80 showed
259 no significant effects on BLIS production.

260

261 The presence of Tween 80 (polyoxyethylene sorbitan mono-oleate) enhanced
262 the production of some bacteriocins by preventing aggregation of bacteriocins
263 molecules (Nissen-Meyer *et al.*, 1992; Parente and Hill, 1992; Moretro *et al.*, 2000).
264 This could be brought about by the formation of micelles in the medium which
265 stabilized the production of bacteriocin ²¹ and facilitated the discharge of bacteriocin
266 from the cell surface ²². However, the effect of Tween 80 is greatly dependent on the
267 type of bacteriocin and the producing strain. As observed in the present study, Tween
268 80 used in the half factorial design did not affect BLIS production which was
269 similarly reported for sakacin A ²³, pediocin A ²⁴, lactocin S ²⁵, plantaricin 423 ²⁶ and
270 pediocin AcH ^{27, 28}.

271

272 Although the buffering capacity of the system due to sodium β-
273 glycerophosphate influenced the growth of bacterium, its role to exclude the effect of
274 organic acids was not significant from the result of this study. The possible
275 explanation on this could be the BLIS from *P. acidilactici* Kp10 was stable at acidic

276 pH¹⁶ and reduced BLIS activity during the fermentation could not be explained by
277 the reduction in the culture pH. The enzymatic reactions are controlled by pH and the
278 decrease in culture pH results to a decrease in enzymatic reactions with concomitant
279 reduction in growth rate.

280

281 Among the three significant factors analyzed, temperature appeared to exhibit
282 the most pronounced effect compared to agitation speed and inoculum size.
283 Temperature plays an important role in bacteriocin production²⁸ where elevated
284 temperature may completely suppress bacteriocin synthesis and sometimes leads to an
285 irreversible loss of its property. The influence of temperature on bacteriocin
286 production was strain dependent. Most of the strains isolated to date, require a
287 temperature ranging from 30 °C to 37 °C for bacteriocin production²⁹. At high
288 temperatures (44 °C), microorganisms were unable to synthesize or secrete
289 bacteriocin and degradation or inactivation could not be increased³⁰. Conversely,
290 high cell yield and high bacteriocin activities at low temperatures could be due to
291 different rate-limiting reactions which are temperature dependent. This resulted in
292 better utilization of carbon and/or energy at low growth rates and increased
293 availability of essential metabolites (ATP included) for bacteriocin synthesis³¹.

294

295 Multilayer perceptron networks (MPN) with back-propagation and conjugate
296 gradient descent (CGD) logarithms were used in the training of neural network on the
297 basis of varying input/output pair data sets. The topology of the network consisted of
298 three layers (3:9:1)- an input layer consisting of three variables, middle hidden layer
299 of 9 neurons and one output layer was used for the prediction and optimization of
300 BLIS production. The neurons activation for ANN processing is represented by

301 different colors (Figure 1). The optimum levels predicted from response graphs, made
302 by ANN for temperature, agitation and inoculum size were 28.5 °C, 120 rpm and 3%
303 respectively. The highest effect was from the temperature, followed by agitation as
304 calculated by the sensitivity analysis (Table 6). From the sensitivity analysis
305 conducted in this study, temperature (ratio = 136.8833) gave the highest effect on
306 BLIS production by *P. acidilactici* Kp10. The ratio parameter described the behavior
307 of the neural network when a variable was removed from the input data. The
308 interaction effects were represented by surface plots (Figure 2). The surface plot
309 between temperature and inoculum size obtained from ANN is as shown in Figure 2a.
310 However, the surface plot between temperature and agitation (Figure 2b) is portrayed
311 as a combination of maximum hill and stationary ridge while the surface plot between
312 inoculum size and agitation (Figure 2c) is depicted as a combination of saddle and
313 stationary ridge.

314

315 Before optimization, maximum BLIS activity obtained experimentally was
316 888.56 AU/mL. Following optimization BLIS activity increased to 5,118.49 AU/mL.
317 The response surface curves were plotted to understand the interaction of the
318 variables and to determine the optimum level of each variable for maximum response
319 ³². The response surface and contours of ANN models for BLIS production are shown
320 in Figure 2 (a, b, c). From the data obtained, it was observed that the lower and higher
321 levels of all variables did not result in higher BLIS production. The shape of the
322 response surface curves showed a moderate interaction between the variables.
323 Comparison of BLIS production by *P. acidilactici* Kp10 before and after optimization
324 and the verification experiments including ANN-predicted optimum levels for the
325 tested variables are presented in Table 7. ANN predicted activity of 5,262.64 AU/mL

326 was obtained at 120 rpm, 3% and 28.5 °C. Experiments in triplicate were conducted
327 on these optimum levels to calculate the observed response. The observed BLIS
328 activity at the predicted optimum levels of the tested variables at ANN predicted
329 levels was 5,118.5. The BLIS activity predicted by ANN was close to the observed
330 BLIS activity at all experimental levels of three variables. ANN results showed a high
331 values of R^2 (1) and adjusted R^2 (1) and low values of AAD (0.49) and RMSE (0.02).

332

333 Analysis of experimental data using ANOVA provided the statistical
334 relationships of the output. The probability decreases as the value of the F value
335 increases. If this probability is less than 0.05 the terms are significant and their
336 inclusion improves the model³³. The present study showed that the effect of
337 temperature (F value: 46.419) was more pronounced than the effect of agitation (F
338 value: 6.419) where both have the probability of less than 0.05. P -values were used as
339 a tool to check significance of each variable which also showed the interaction
340 strength between each independent variable³⁴. The smaller the P -values, the higher is
341 the significance of the corresponding variable³³. Larger values of RMSE and AAD
342 mean higher chances of errors in prediction. In this study, lower RMSE and ADD
343 values obtained from ANN were more reliable and accurate. Experimental results
344 from ANN were in good agreement with the predicted values indicating that actual
345 and predicted values confirmed each other and that the models were reasonable and of
346 high accuracy in predicting the values of the dependent variables.

347

348 **Conclusions**

349 ANN provided a confident estimation capability through the range of variables in the
350 optimization of BLIS production by *P. acidilactici* Kp10. The ability of ANN to

351 predict the process characteristics with little prior knowledge is desirable which
352 simplify its implementation and increase its modeling potential. This property makes
353 ANN a powerful and flexible tool well-suited to modeling of complex bioprocesses.
354 As observed from the present study, BLIS production by *P. acidilactici* Kp10 in
355 optimized fermentation (5,118.49 AU/mL) was about 6 times higher than that
356 obtained in non-optimized fermentation (888.56 AU/mL).

357

358 **Competing interests**

359 The authors declare that they have no competing interests.

360

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Figure Captions

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Fig. 1

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The topology of neural network for estimation of BLIS production by *Pediococcus acidilactici* Kp10. Triangles represent the inputs (Neurons added for ANN processing); inoculum size temperature and agitation. Squares represent the hidden and output layer (neurons generated during ANN processing). Small open circles indicate the input and output layers (the neurons that can be observed in the form of numerical values). Colours indicate the activation level of neurons; (red) positive activation; (green) negative activation. Intensity of colour represents the activation intensity for ANN processing.

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Fig. 2

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Surface plot obtained from optimization using ANN for the combined effect of (a) temperature and inoculum size (by keeping agitation at center point); (b) temperature and agitation (by keeping inoculum size at center point); (c) inoculum size and agitation (by keeping temperature at center point) on BLIS production by *Pediococcus acidilactici* Kp10. The different colours in the legend represent the respective range of BLIS production.

Table 1 Half factorial screening experimental design used to test the effect of several fermentation parameters on the production of BLIS by *P. acidilactici* Kp10.

Run no.	Temp (X ₁)	Inoc. size (X ₂)	Buffer strength (X ₃)	Tween 80 conc. (X ₄)	Agit. speed (X ₅)	Response		
	(°c)	%	µM	g/L	rpm	OD (650 nm)	Activity (AU/mL)	pH
1	1(37)	-1(1)	1(180)	-1(0.16)	1(200)	1.81±0	573.55	6.5
2	1(37)	1(10)	1(180)	-1(0.16)	-1(0)	0.5±0.04	1738.58	6.5
3	1(37)	1(10)	1(180)	1(1.6)	1(200)	1.59±0	767.92	6.7
4	-1(20)	1(10)	1(180)	1(1.6)	-1(0)	0.89±0	1028.15	6.8
5	-1(20)	1(10)	-1(60)	-1(0.16)	-1(0)	0.84±0.16	1897.65	6.3
6	1(37)	-1(1)	-1(60)	1(1.6)	1(200)	2.44±0	1417.34	6.0
7	-1(20)	1(10)	1(180)	-1(0.16)	1(200)	3.31±0.02	3208.88	6.3
8	1(37)	1(10)	-1(60)	-1(0.16)	1(200)	1.61±0	428.38	6.1
9	-1(20)	1(10)	-1(60)	1(1.6)	1(200)	3.51±0.06	3822.95	6.6
10	-1(20)	-1(1)	1(180)	-1(0.16)	-1(0)	0.73±0.07	663.66	7.0
11	-1(20)	-1(1)	1(180)	1(1.6)	1(200)	1.58±0	1738.58	7.01
12	-1(20)	-1(1)	-1(60)	-1(0.16)	1(200)	3.19±0.3	1592.83	6.3
13	1(37)	-1(1)	-1(60)	-1(0.16)	-1(0)	0.95±0.02	1122.22	6.1
14	-1(20)	-1(1)	-1(60)	1(1.6)	-1(0)	0.83±0.01	863.00	6.5
15	1(37)	-1(1)	1(180)	1(1.6)	-1(0)	0.33±0.01	1089.95	6.5
16	1(37)	1(10)	-1(60)	1(1.6)	0	0.84±0.01	1189.67	6.1

±: Standard deviation of duplicate data

Value of factors in pranthesis are actual level with respect to the coded values.

Table 2 Analysis of variance (ANOVA) for the screening of BLIS production by *Pediococcus acidilactici* Kp10.

Source	Coefficient estimate	Standard Error	Sum of squares	DF	Mean square	F Value	Prob>F
Model			1.282E+007	13	9.860E+005	109.32	0.0091
Intercept	1446.46	23.74					
<i>A</i>	-405.51	23.74	2.631E+006	1	2.631E+006	291.70	0.0034
<i>B</i>	313.82	23.74	1.576E+006	1	1.576E+006	174.70	0.0057
<i>C</i>	-95.30	23.74	1.453E+005	1	1.453E+005	16.11	0.0568
<i>D</i>	43.24	23.74	2991.57	1	2991.57	3.32	0.2102
<i>E</i>	247.35	23.74	9.789E+005	1	9.789E+005	108.53	0.0091
<i>AB</i>	-323.63	23.74	1.676E+006	1	1.676E+006	185.80	0.0053
<i>AC</i>	96.85	23.74	1.501E+005	1	1.501E+005	16.64	0.0552
<i>AD</i>	32.03	23.74	16415.38	1	16415.38	1.82	0.3098
<i>AE</i>	-491.50	23.74	3.865E+006	1	3.865E+006	428.54	0.0023
<i>BD</i>	-101.34	23.74	1.643E+005	1	1.643E+005	18.22	0.0508
<i>BE</i>	49.41	23.74	39066.51	1	39066.51	4.33	0.1729
<i>CD</i>	-238.25	23.74	9.082E+005	1	9.082E+005	100.69	0.0098
<i>DE</i>	199.66	23.74	6.378E+005	1	6.378E+005	70.71	0.0138
Residual			18038.83	2	9019.41		
Cor Total			1.284E+007	15			

A: temperature

B: inoculum size

C: buffer strength

D: tween 80

E: agitation

italic= significant term

Table 3 Box-Wilson 2^3 factorial central composite design for optimization of BLIS production by *P.acidilactici* using ANN.

Exp. No.				AU/mL		Final culture pH
	Temperature (X ₁) (°C)	Inoculum Size (X ₂)(%)	Agitation speed (X ₃) (rpm)	Observed	Predicted by ANN (% difference ^a)	
1	- α (20)	0(5.5)	0(100)	888.6	879.1(-1.07)	6.4
2	-α(20)	0(5.5)	0(100)	888.6	879.1(-1.07)	6.4
3	-1(23.45)	+1(8.18)	+1(159.46)	1592.8	1586.2(-0.42)	6.24
4	-1(23.45)	+1(8.18)	-1(40.54)	888.6	923.2(3.90)	6.53
5	-1(23.45)	-1(2.82)	-1(40.54)	1592.8	1577.3(-0.97)	6.56
6	-1(23.45)	+1(8.18)	-1(40.54)	888.6	923.2(3.90)	6.54
7	-1(23.45)	+1(8.18)	+1(159.46)	1592.8	1586.1(-0.42)	6.27
8	-1(23.45)	-1(2.82)	+1(159.46)	2855.3	2876.2(0.73)	6.19
9	-1(23.45)	-1(2.82)	+1(159.46)	2855.3	2876.2(0.73)	6.18
10	-1(23.45)	-1(2.82)	-1(40.54)	1592.8	1577.3(-0.97)	6.55
11	0(28.5)	+ α (10)	0(100)	5118.5	5096.2(-0.44)	6.5
12	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.41
13	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.41
14	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.41
15	0(28.5)	0(5.5)	-α(0)	5118.5	5098.9(-0.38)	6.27
16	0(28.5)	0(5.5)	+ α (200)	1592.8	1587.0(-0.35)	6.24
17	0(28.5)	0(5.5)	+α(200)	1592.8	1587.0(-0.35)	6.21
18	0(28.5)	- α (1)	0(100)	5118.5	5135.0(0.32)	6.36
19	0(28.5)	-α(1)	0(100)	5118.5	5135.0(0.32)	6.32
20	0(28.5)	+α(10)	0(100)	5118.5	5096.2(-0.44)	6.49
21	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.43
22	0(28.5)	0(5.5)	- α (0)	5118.5	5098.9(-0.38)	6.26
23	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.45
24	0(28.5)	0(5.5)	0(100)	5118.5	5123.8(0.10)	6.41

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25	+1(33.55)	+1(8.18)	+1(159.46)	1592.8	1619.9(1.70)	6.29
26	+1(33.55)	+1(8.18)	+1(159.46)	1592.8	1619.9(1.70)	6.29
27	+1(33.55)	-1(2.82)	+1(159.46)	1592.8	1582.9(-0.62)	6.3
28	+1(33.55)	-1(2.82)	-1(40.54)	1592.8	1604.3(0.72)	6.41
29	+1(33.55)	-1(2.82)	+1(159.46)	1592.8	1582.9(-0.62)	6.33
30	+1(33.55)	+1(8.18)	-1(40.54)	1592.8	1593.8(0.06)	6.41
<i>31</i>	<i>+1(33.55)</i>	<i>+1(8.18)</i>	<i>-1(40.54)</i>	<i>1592.8</i>	<i>1593.8(0.06)</i>	<i>6.41</i>
<i>32</i>	<i>+1(33.55)</i>	<i>-1(2.82)</i>	<i>-1(40.54)</i>	<i>1592.8</i>	<i>1604.3(0.72)</i>	<i>6.4</i>
33	+ α (37)	0(5.5)	0(100)	2132.6	2122.0(-0.50)	6.4
34	+α(37)	0(5.5)	0(100)	2132.6	2122.0(-0.50)	6.4

The italic, bold and normal values represent the experiments used for selection, training and testing, respectively, by the selected ANN.

^a% difference was calculated as the % difference between the observed value and corresponding predicted value over the observed value.

Table 4 Effect of two types of medium and cultivation temperature on growth of *P. acidilactici* Kp10 and BLIS production.

Fermentation time (h)	30 °C				35 °C				37 °C			
	MRS		M17		MRS		M17		MRS		M17	
	OD (650 nm)	AU/mL	OD (650 nm)	AU/mL	OD (650 nm)	AU/mL	OD (650 nm)	AU/mL	OD (650 nm)	AU/mL	OD (650 nm)	AU/mL
0	0.07	57.6	0.08	103.3	0.07	57.6	0.08	103.3	0.07	57.6	0.08	103.3
3	0.09	62.1	0.23	103.3	0.08	57.6	0.25	214.2	0.09	57.6	0.3	103.3
6	0.09	66.7	0.60	220.5	0.09	69.8	0.72	331.8	0.09	65.8	0.69	322.3
9	0.09	83.2	0.67	545.0	0.09	72.8	0.73	594.8	0.09	72.8	0.79	594.8
12	0.09	124.3	0.72	921.5	0.09	75.1	0.77	1035.6	0.09	73.0	0.72	688.3
15	0.09	160.0	0.75	1386.6	0.09	90.0	0.81	1035.6	0.09	74.9	0.72	921.5
18	0.09	138.3	0.75	1427.7	0.09	83.2	0.78	1187.2	0.08	43.0	0.71	854.2
21	0.09	90.1	0.75	967.2	0.09	65.4	0.7	765.3	0.08	36.2	0.69	796.4
24	0.08	43.0	0.62	456.3	0.09	37.2	0.66	523.1	0.08	32.1	0.66	624.1

Size of inoculum: 10% (v/v)

Initial pH for:

1) **M17 broth:** 7.2 ± 0.2

2) **MRS broth:** 5.7 ± 0.2

Table 5 Analysis of variance for BLIS optimization using Box-Wilson 2^3 factorial central composite design.

Variables	Analysis of variance				Parameter estimates				
	SS	DF	MS	F	Estimates	t values	P values	Confidence limits	
								-95%	+95%
Intercept	39150376	1	39150376	33.1865	-50107.3	-5.7607	0.0000	-68059.1	-32155.5
Temperature (X₁)	54761709	1	54761709	46.4197	3621.8	6.8132	0.0000	2524.6	4718.9
Temperature² (X₁²)	59264102	1	59264102	50.2362	-63.5	-7.0877	0.0000	-82.0	-45.0
Inoculum size (X ₂)	23	1	23	0.0000	3.1	0.0043	0.9965	-1436.0	1442.1
Inoculum size ² (X ₂ ²)	2694169	1	2694169	2.2837	-48.3	-1.5112	0.1437	-114.2	17.7
Agitation (X₅)	7254150	1	7254150	6.1491	75.7	2.4797	0.0205	12.7	138.6
Agitation² (X₅²)	21167884	1	21167884	17.9433	-0.3	-4.2359	0.0002	-0.4	-0.1
Temperature(X ₁) * Inoculum size(X ₂)	966977	1	966977	0.8196	18.2	0.9053	0.3742	-23.3	59.6
Temperature(X ₁) * Agitation(X ₅)	966977	1	966977	0.8196	-0.8	-0.9053	0.3742	-2.7	1.0
Inoculum size (X ₂) * Agitation(X ₅)	77925	1	77925	0.0660	-0.4	-0.2570	0.7993	-4.0	3.1
Error	28312989	24	1179708						

Bold letters represented significant variables and their calculated values. SS stands for sum of squares and MS stands for mean sum of squares.

1 **Table 6** Sensitivity Analysis by ANN.

Parameter	Temperature (X₁) (°C)	Inoculum Size (X₂)(%)	Agitation speed (X₅) (rpm)
Ratio	136.88	15.91	70.1
Rank	1	3	2

2 Ratios are values given by ANN as a result of sensitivity analysis to input variables.
 3 Rank is the order according to the ratios.

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39 **Table 7** Predicted optimal levels by ANN and observed and predicted response on
 40 these levels.

Sr. No.	Method	Agitation Speed (rpm)(X ₅)	Inoculum Size (%) (X ₂)	Temperature (°C)(X ₁)	BLIS production (Y)(AU/mL)	
					Predicted by ANN	Experimental
1	Base case	0	10	37	1,912.2	888.6
2	ANN predicted optima	120	3	28.5	5,262.6	5,118.5

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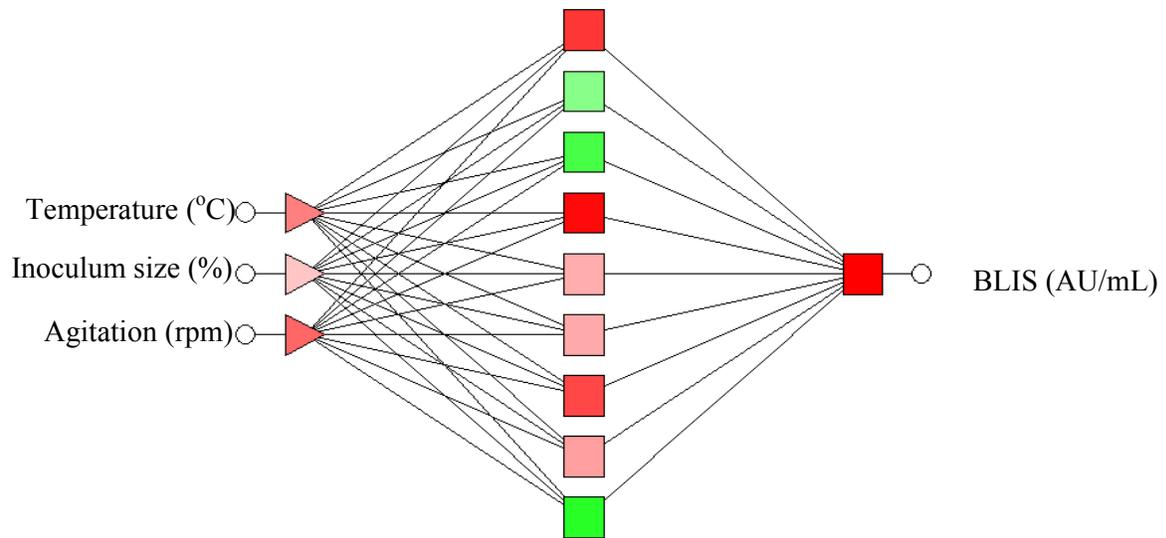


Fig. 1.

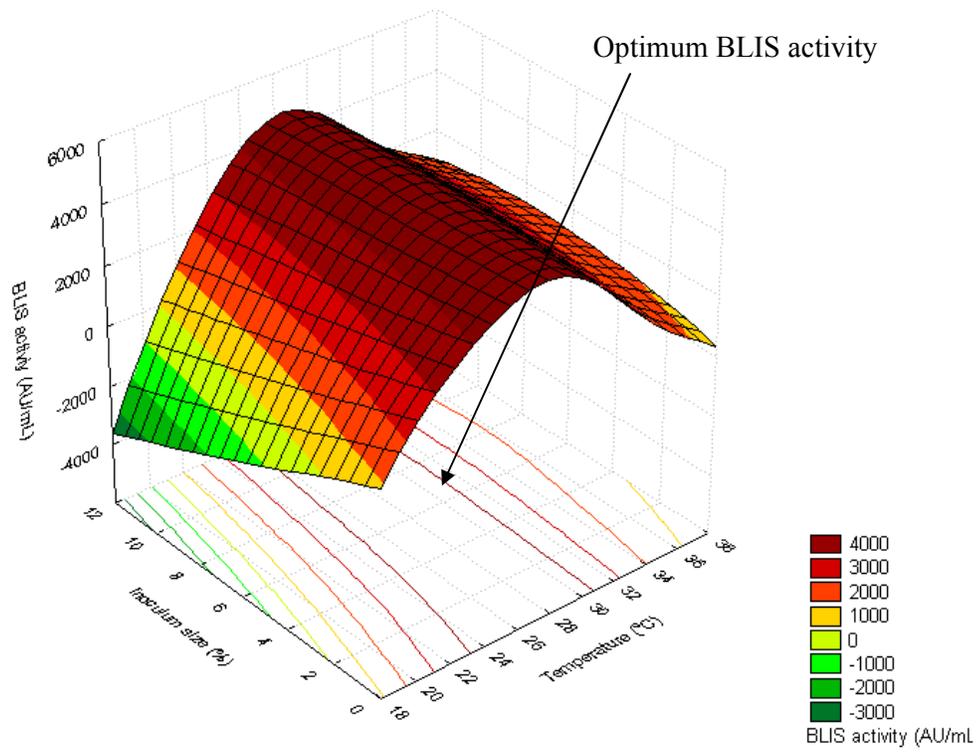


Fig. 2a.

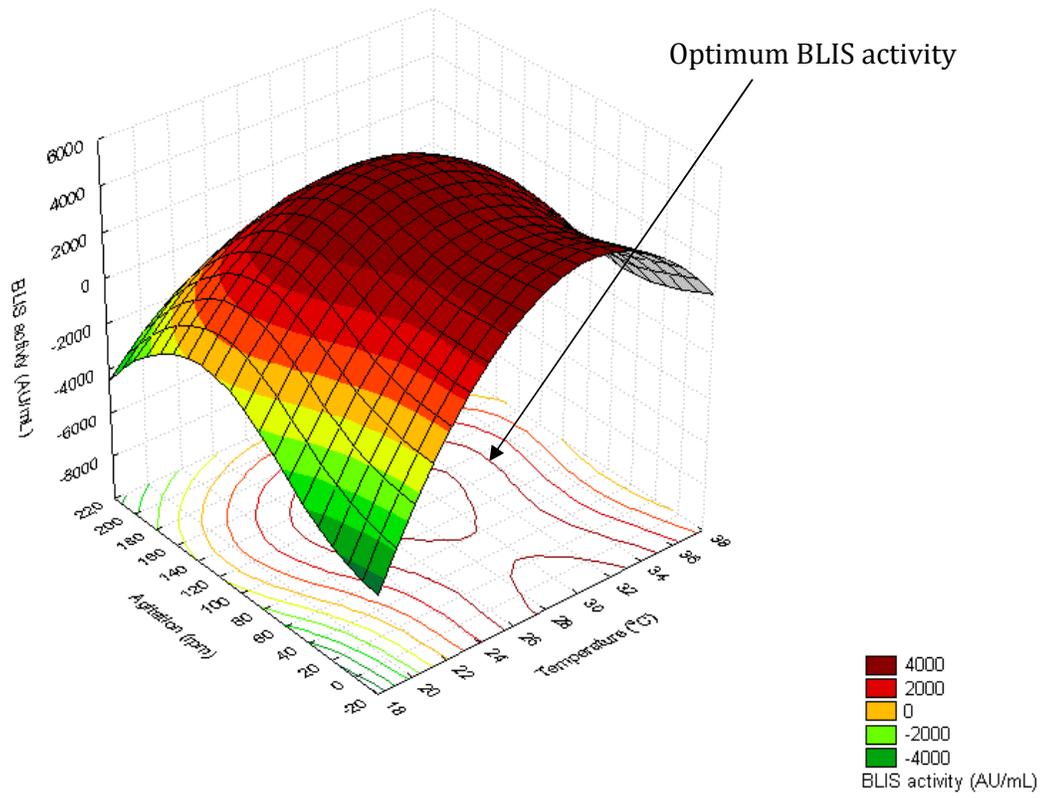
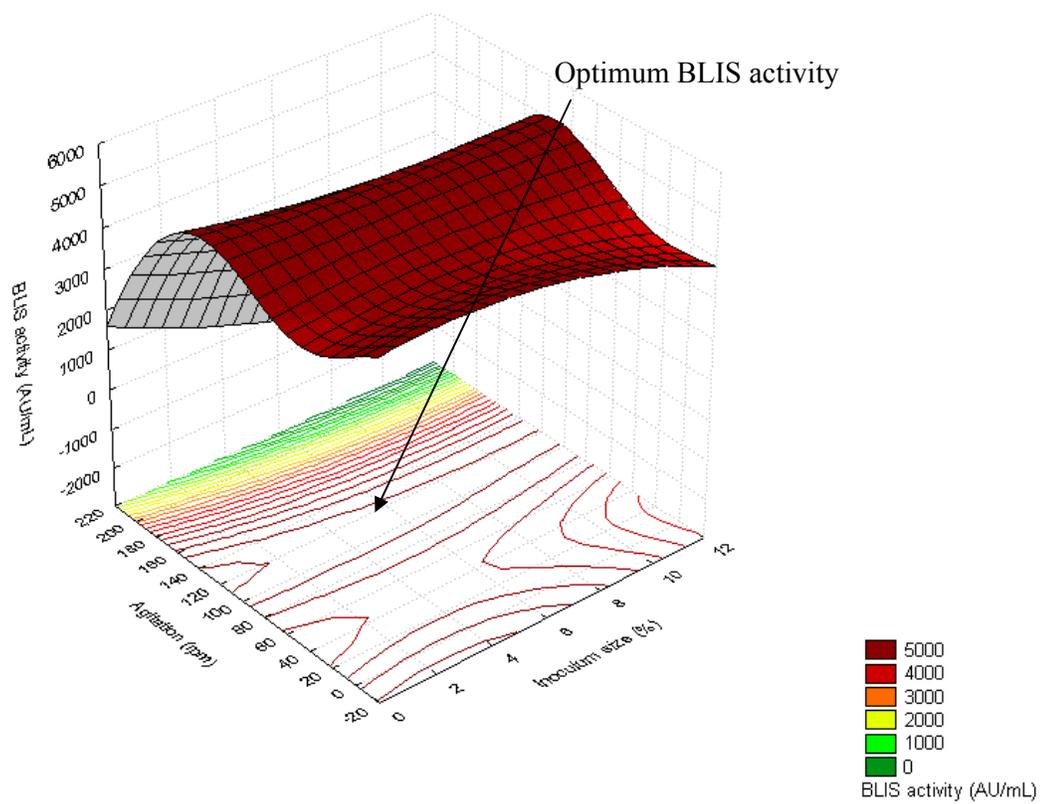


Fig. 2b.

**Fig. 2C.**