

# Journal Pre-proof

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PII: S1878-8181(21)00112-2

DOI: <https://doi.org/10.1016/j.bcab.2021.102016>

Reference: BCAB 102016

To appear in: *Biocatalysis and Agricultural Biotechnology*

Received Date: 23 February 2021

Revised Date: 23 April 2021

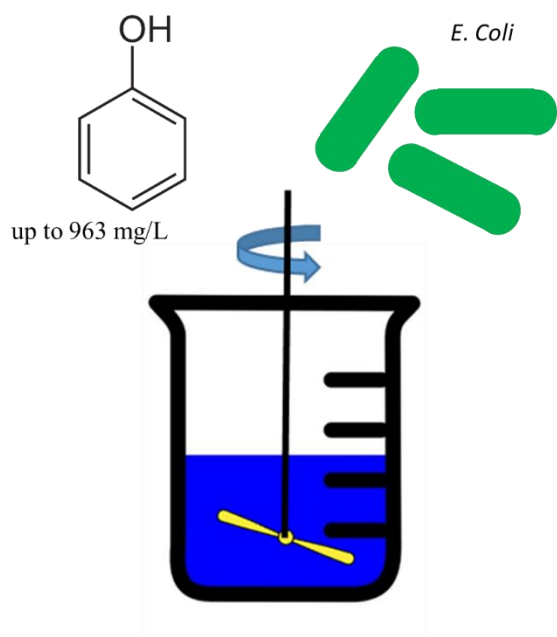
Accepted Date: 26 April 2021

Please cite this article as: Bouchene, H., Mahy, J.G., Lambert, S.D., Hayoun, B., Deflaoui, O., Bourouina, M., Bachari, K., Hamitouche, A.-E., Bacha-Bourouina, S., Statistical modeling and optimization of *Escherichia coli* growth parameters for the biological treatment of phenol, *Biocatalysis and Agricultural Biotechnology*, <https://doi.org/10.1016/j.bcab.2021.102016>.

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**Biodegradation process optimization  
thanks to Baranyi and Roberts'  
equations :**

Study of pH, initial concentrations of ammonium sulfate and phenol on growth parameters: latency time,  $\lambda$  (h), maximum specific growth rate,  $\mu_{\max}$  ( $\text{h}^{-1}$ ), and maximum optical density,  $\text{OD}_{\max}$

**Optimized parameters:**

$\lambda_{\min} = 21.08$  h,  $\mu_{\max} = 8.68$   $\text{h}^{-1}$  and  $\text{OD}_{\max} = 0.39$  at  $\text{pH} = 6.3$  for  $[\text{phenol}]_0 = 200$  mg/L and  $[(\text{NH}_4)_2\text{SO}_4]_0 = 1.33$  g/L

1 **Statistical modeling and optimization of *Escherichia coli* growth parameters for**  
2 **the biological treatment of phenol**

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19

20

21 **Abstract**

22 Aromatic compounds, including phenols, are a significant source of pollution which need to  
23 be treated by environmentally-friendly methods, such as bioprocesses. This work focuses on  
24 the biodegradation of phenol in a batch reactor with bacteria, and the optimization of the  
25 growth parameters in order to obtain the highest phenol degradation. The model and

26 algorithms fitting the growth data are emphasized. Primary models, applied to monitor the  
27 dynamic evolution of the microbial biomass of the selected strain, were fitted to the data by  
28 nonlinear regression based on the Levenberg Marquart algorithm. The statistically-validated  
29 Baranyi and Roberts equation was used to evaluate the growth parameters: maximum growth  
30 rate ( $\mu_{\max}$ ), latency time ( $\lambda$ ), and maximum optical density ( $OD_{\max}$ ). To improve bacterial  
31 growth and phenol degradation performance, physico-chemical conditions, such as initial  
32 phenol concentration, pH, and nitrogen source (ammonium sulfate), were optimized using  
33 secondary models based on a central composite rotatable design (CCRD). The correlation  
34 coefficient,  $R^2$ , for each regression equation is  $> 94\%$ . The optimal values of growth  
35 parameters are  $\lambda_{\min} = 21.08$  h,  $\mu_{\max} = 8.68$  h<sup>-1</sup>, and  $OD_{\max} = 0.39$  at pH = 6.3 for an initial  
36 concentration of phenol = 200 mg/L and initial concentration of ammonium sulfate =  
37 1.33 g/L. *Escherichia coli* showed an ability to degrade up to 963 mg/L of phenol in 250 h  
38 without prior acclimatization of the strain.

39

40 **Keywords:** Phenol biodegradation; *Escherichia coli*; statistical modeling; Baranyi and  
41 Roberts model; central composite rotatable design.

42

### 43 1. Introduction

44 The treatment of environmental pollution is one of the major concerns of the 21<sup>st</sup> century,  
45 especially water pollution which has been of great importance for the last four decades.  
46 Aromatic compounds, including phenolic compounds, are known for their omnipresence in  
47 industrial effluents.

48 The toxicity of these products requires their elimination from water before it is discharged  
49 into rivers (Nair et al., 2008). Phenol is a basic reagent (Basha et al., 2010) used in numerous  
50 processes, such as coal processing plants (Pradeep et al., 2015; Zhai et al., 2012), petroleum

51 refineries, pulp and paper mills, iron and steel industries, plastics and varnishes, textile mills,  
52 pesticide factories, companies producing polymer resins, pharmaceutical industries, fiberglass  
53 manufacturing, and wood preservation plants (Al-Khalid and El-Naas, 2012; Iqbal et al.,  
54 2017; Ke et al., 2018; Zou et al., 2018). Acute exposure to phenol is known to cause  
55 gastrointestinal discomfort and headaches (Pradeep et al., 2015). Also, phenol is toxic to the  
56 nervous system, heart, kidneys, and liver and is easily absorbed through the skin (Zhai et al.,  
57 2012). Similarly, for aquatic organisms, phenol and its derivatives are toxic or lethal at  
58 concentrations from 5 to 25 mg/L (Al-Khalid and El-Naas, 2012).

59 Because of this toxicity, the Algerian regulations in accordance with Executive Decree No.  
60 09-209 of June 11, 2009, describing the procedure for granting authorization to discharge  
61 non-domestic wastewater into a public sewerage network or a wastewater treatment plant, set  
62 1 mg/L as the maximum limit for phenol content in this non-domestic wastewater. Also,  
63 Algerian Executive Decree No. 06-141 of April 19, 2006, defining the limit values for  
64 discharges of industrial liquid effluents, stipulated a value of 0.3 mg/L of effluents from the  
65 energy industry. Finally, the World Health Organization has established a guideline of 1 µg/L  
66 to regulate the concentration of phenol in drinking water, and a more severe guideline of  
67 0.5 µg/L was set by the European Council's Drinking Water Directive. Moreover, the United  
68 States Environmental Protection Agency has defined phenol as a priority pollutant.

69 To comply with these standards of rejection, several physical or chemical treatment methods  
70 are available, such as activated carbon adsorption, ion exchange, liquid-liquid extraction, or  
71 chemical oxidation (Al-Khalid and El-Naas, 2012). Unfortunately, these methods often suffer  
72 from drawbacks, such as high cost, or they simply transfer the pollution from one phase to  
73 another (e.g., from the liquid water to a solid such as carbon) (Al-Khalid and El-Naas, 2012).

74 In contrast, biological processes (using microorganisms for phenol degradation) are more  
75 attractive because they are considered to be more environmentally friendly (Koyande et al.,

76 2019; Tang et al., 2020; Yew et al., 2021) and more economical than chemical and physical  
77 treatments (Chai et al., 2021; Ke et al., 2018; Pradeep et al., 2015; Rambabu et al., 2020;  
78 Show et al., 2012; Tang et al., 2020; Xie et al., 2021, 2020; Zhai et al., 2012). Several  
79 microorganisms (bacteria, fungi, yeasts, and microalgae) were studied for the biodegradation  
80 of phenol (Basha et al., 2010; Lindner and Pleissner, 2019; Singh and Kumar, 2019). These  
81 microorganisms use the phenol as a carbon source for their development during cell growth,  
82 as phenol is the only source of carbon present in the medium. A biodegradation study cannot  
83 be carried out without optimizing the kinetic parameters of the growth of the microorganisms  
84 responsible for the degradation of phenol in order to ensure good microorganism  
85 development. To highlight some of the relationships between all the different parameters,  
86 statistical experimental designs provide the best methodologies to optimize and evaluate the  
87 factors that influence the biodegradation process of phenol (Presser et al., 1997; Tebbouche et  
88 al., 2015).

89 The aim of this study is to evaluate the capacity for phenol biodegradation of *Escherichia coli*  
90 *RV412\_AI\_AI\_2010\_06a LBK* isolated from the wastewater treatment plant of Beni Messous  
91 (Algeria). First, a central composite rotatable design (CCRD) was chosen to study the  
92 influence of three factors, namely pH, the initial phenol concentration, and initial nitrogen  
93 source concentration on the growth kinetics of the isolated bacteria during phenol  
94 degradation. The bacterial growth parameters, such as maximum growth rate ( $\mu_{\max}$ ), latency  
95 time ( $\lambda$ ), and maximum optical density ( $OD_{\max}$ ), were previously estimated using two primary  
96 models widely applied in microbiology: Gompertz, and Baranyi and Roberts. The polynomial  
97 regression equations are established for these kinetic growth parameters as a function of the  
98 three factors and their interactions. The second part of this work concerns the study of the  
99 influence of the operating factors on the phenol degradation yield and its relation to the

100 evolution of bacterial growth parameters. This bacterial strain can be applied for the treatment  
101 of highly concentrated solutions of phenol (up to 1000 mg/L).

102

## 103 2. Methods

### 104 2.1. Microorganism type and culture conditions

105 The bacterial strain used in this work was collected and isolated from the Wastewater  
106 Treatment Plant of Beni Messous station (Algiers, Algeria). The bacterial culture was stored  
107 at 4°C. For all experiments, the cultured biomass was inoculated into 100 mL of culture  
108 medium for 24 h at 30°C and recovered by centrifugation (3000 rpm for 30 min) (Hamitouche  
109 et al., 2010). The culture medium was composed of peptone (15 g/L), yeast extract (3 g/L),  
110 sodium chloride (6 g/L), and (D+)-glucose (1 g/L). The biomass obtained was rinsed with a  
111 0.5% NaCl solution and centrifuged again. The rinsed microorganisms were then suspended  
112 in 0.5% NaCl (Hamitouche et al., 2010). All cultures were carried out at 1% V/V at an initial  
113  $OD_{600nm} = 0.09$ .

114

### 115 2.2. Identification of the bacteria

116 The identification of the bacteria was achieved by mass spectroscopy (Labella et al., 2021;  
117 Sun et al., 2021) thanks to the matrix assisted laser desorption ionization - time of flight  
118 (MALDI-TOF) technique using a device from Bruker Daltonik (Billerica, MA, USA). The  
119 mass spectrum obtained was compared to a database of reference spectra (RUO MALDI  
120 Biotyper Reference Library v3.3.1.2, SR library v1.0) by the software (Bruker Daltonik) to  
121 determine relatedness to spectra in the database (Shah and Gharbia, 2017).

122 ~~The principle of mass spectroscopy is the detection and identification of molecules according~~  
123 ~~to their mass and the characterization of their chemical structure by gas separation of ions~~  
124 ~~according to their m/Z ratio. MALDI TOF is a technique of ionization of biomolecules. This~~

125 method involves three major steps: (i) deposition of the sample on a metal plate by mixing it  
126 with a matrix which crystallizes it, (ii) ionization by bombardment with brief laser pulses,  
127 producing positively charged analytes, (iii) separation of ions formed by this process  
128 according to their TOF in the flight tube. Ion collisions in the flight tube are avoided by the  
129 generation of a vacuum pushed by a pump, before the laser pulses, which takes 1-2 min.  
130 Protein detection at the exit of flight tube leads to the production of a mass spectrum  
131 composed of specific mass peaks with different intensities that express a reproducible  
132 fingerprint of a microorganism. At the end of the analysis, the spectra obtained are compared  
133 with reference databases (Shah and Gharbia, 2017).

134

### 135 2.3. Biodegradation of phenol

136 The biodegradation experiments were carried out under aseptic conditions, 1 mL of the  
137 conserved culture was inoculated into 100 mL of Nutrient Broth Medium and incubated for  
138 24 h at 30°C. The resulting biomass was suspended in NaCl 0.5% solution as previously  
139 described. The initial density was measured by spectrophotometry at 600 nm, and was equal  
140 to 2.5 to 2.9. When inoculated into a 250 mL Erlenmeyer flask containing 100 mL saline  
141 medium,  $OD_{600\text{ nm}} = 0.09$ .

142 The composition of the culture medium was (g/L):  $\text{NaH}_2\text{PO}_4$ : 3,  $\text{KH}_2\text{PO}_4$ : 3,  $\text{MgSO}_4$ : 0.1, and  
143 the nitrogen source ( $(\text{NH}_4)_2\text{SO}_4$ ) was optimized in the range of 0 to 2 g/L, the carbon source  
144 (phenol) varied in the range of 50 to 200 mg/L. pH was optimized within the range of 6 to 8.  
145 All experiments were performed at 30°C. Similarly, the effect of the initial phenol  
146 concentration (65, 562, and 963 mg/L) on microorganism growth and phenol degradation was  
147 examined.

148

## 149 2.4. Design of the experiments

150 The modeling and optimization of the responses were carried out using a central composite  
 151 rotatable design (CCRD) provided by the software Minitab 17. This design was chosen to  
 152 evaluate the main effects and the interactions of three parameters: the initial concentration of  
 153 phenol (50 to 200 mg/L),  $[phenol]_0$ , the initial concentration of ammonium sulfate (0 to 2  
 154 g/L),  $[(NH_4)_2SO_4]_0$ , and pH (6 to 8), on three responses: the latency time,  $\lambda$  (h), the maximum  
 155 specific growth,  $\mu_{max}$  ( $h^{-1}$ ), and the maximum optical density,  $OD_{max}$ . The statistical  
 156 significance of these models was evaluated using analysis of variance (ANOVA).

157 The correlations between the operating variables (pH,  $[(NH_4)_2SO_4]_0$ , and  $[phenol]_0$ ) and the  
 158 responses were estimated by a full quadratic model (Eq. 1):

159

$$160 Y = b_0 + b_1 x_1 + b_2 x_2 + b_3 x_3 + b_{12} x_1 x_2 + b_{13} x_1 x_3 + b_{23} x_2 x_3 + b_{11} x_1^2 + b_{22} x_2^2 + b_{33} x_3^2 + \dots$$

161 (Eq. 1)

162

163 where  $b_0$  is constant and is defined as the response mean,  $b_j$  represents the linear effects with  
 164  $i=1, 2, \text{ or } 3$ , and  $b_{ij}$  represents the interaction effects with  $i=1, 2, \text{ or } 3$  and  $j=1, 2, \text{ or } 3$ .

165 The growth parameters (the latency time,  $\lambda$  (h), the maximum specific growth,  $\mu_{max}$  ( $h^{-1}$ ), and  
 166 the maximum optical density,  $OD_{max}$ ) were determined for each iteration of the design using  
 167 the most appropriate primary model.

168

## 169 2.5. Kinetic models of growth

170 Growth curves were simulated using two primary models, the modified Gompertz model  
 171 (Eq. 2) and the Baranyi and Roberts model (Eq. 3) (Imadalou-Idres et al., 2018; Romano et  
 172 al., 2007). The cell quantification was expressed in terms of optical density at 600 nm,

173 analyzed by spectrophotometry (Baranyi et al., 1993; Lopez et al., 2004; Zwietering et al.,  
174 1990):

175

$$176 \quad \ln N = \ln N_0 + A e^{-e^{\left\{\left(\frac{\mu_{max} e^1}{A}\right)^{(\lambda-1)}\right\}+1}} \quad (\text{Eq. 2})$$

$$177 \quad \ln N = \ln N_0 + \mu_{max} t + \ln[e^{-\mu_{max} t} - e^{-\mu_{max}(t+\lambda)} + e^{-\mu_{max}\lambda}] - \ln\left[1 + \frac{(e^{\mu_{max}(t-\lambda)} + e^{-\mu_{max}\lambda})}{e^{(\ln N_{max} - \ln N_0)}}\right] \quad (\text{Eq. 3})$$

178

179 where  $N$  denotes the microbial biomass measured at time  $t$ ;  $N_0$  and  $N_{max}$  are the microbial  
180 biomass measured at  $t = 0$  and at maximum microbial population size, respectively;  $A$  is the  
181 asymptotic level =  $\ln\left(\frac{N_{max}}{N_0}\right)$ ;  $\mu_{max}$  is the maximum specific growth rate ( $\text{h}^{-1}$ ), and is defined as  
182 the tangent to the inflection point of the growth curve;  $\lambda$  is the latency phase duration (h);  $t$  is  
183 time (h).

184

## 185 2.6. Analytical methods

186 The residual phenol was quantified by applying the colorimetric method by complexation of  
187 phenol with potassium ferricyanide (Yang and Humphrey, 1975). For the colorimetric  
188 quantification of phenol, 1.15 mL NaAc-HAc buffer (0.2 M, pH 4.0) and 100  $\mu\text{L}$  potassium  
189 ferricyanide ( $0.5 \text{ mg mL}^{-1}$ ) were first mixed together, and then 900  $\mu\text{L}$  phenol at different  
190 concentrations was added to the above solution (Yang and Humphrey, 1975). After incubation  
191 at room temperature for 30 min, the system was measured by using a visible  
192 spectrophotometer Hamatsu at 510 nm (Rodier and Legube, 2009).

193

## 194 3. Results and discussion

### 195 3.1. Identification of biomass

196 In this study, the bacterial strain used was isolated from a wastewater treatment plant. Before  
197 proceeding with the modeling and optimization of biodegradation processes, it is necessary to  
198 start by identifying the isolated strain using MALDI-TOF spectroscopy, a method based on  
199 the analysis of the protein composition produced by the analyzed sample (Asakawa et al.,  
200 2013).

201 Two reproducible MALDI-TOF spectra were obtained for two different samples (Figure 1).  
202 Comparison of these spectra with the database of reference spectra (RUO MALDI Biotyper  
203 Reference Library v3.3.1.2, SR library v1.0) revealed that the bacterial strain is an  
204 *Escherichia coli* RV412\_AI\_2010\_06a LBK with a value equal to 2.259 (Schoch et al., 2020).  
205 The closer this score value is to 3, the higher the probability of species identification  
206 accuracy.

207

### 208 3.2. Kinetic model

209 The curve of Figure 2 shows the growth of *E. coli* bacteria in aerobic conditions at 30°C and  
210 medium containing (g/L): NaH<sub>2</sub>PO<sub>4</sub>: 3, KH<sub>2</sub>PO<sub>4</sub>: 3, MgSO<sub>4</sub>: 0.1, (NH<sub>4</sub>)<sub>2</sub>SO<sub>4</sub>: 1, and phenol as  
211 the sole source of carbon at an initial concentration equal to 125 mg/L. Three distinct growth  
212 phases are observed: (i) the latency period, which serves to adapt the inoculum ( $N_0$ ) to its new  
213 environment. During this period, the specific growth rate is equal to zero ( $\mu = 0$ ); (ii) the  
214 exponential phase where specific growth rate of the bacteria is maximum ( $\mu = \mu_{\max}$ ); (iii) the  
215 stationary period, for which the growth rate is equal to zero ( $\mu = 0$ ) and the culture reaches its  
216 maximum density ( $N = N_{\max}$ ) (Imadalou-Idres et al., 2018).

217 The growth curves obtained for the various operating conditions follow the models currently  
218 applied in microbiology, such as the Gompertz model (Imadalou-Idres et al., 2018) or Baranyi  
219 and Roberts model (Romano et al., 2007). The parameters of these models were estimated by

220 fitting curves to experimental growth data by nonlinear regression (Minitab 17). Both models  
221 have previously been successfully applied to the modeling of *Escherichia coli* growth curves  
222 (Fujikawa et al., 2004; Lopez et al., 2004).

223 Table 1 represents the analysis of variance for the Gompertz and Baranyi and Roberts primary  
224 models. The values of coefficients  $R^2$  and RMSE are very similar between both models. As  
225 the Baranyi and Roberts model presents more advantageous statistical properties compared to  
226 the Gompertz model, the Baranyi and Roberts model was chosen to calculate the kinetic  
227 parameters of bacterial growth.

228

### 229 3.3. Central composite rotatable design evaluation

230 The bacteria used in this work showed an ability to degrade phenol. To understand this  
231 process, three parameters influencing growth kinetics were studied: pH of the solution, the  
232 initial concentration of ammonium sulfate,  $[(NH_4)_2SO_4]_0$ , and the initial concentration of  
233 phenol,  $[phenol]_0$ . The growth parameters, latency time,  $\lambda$ , the maximum specific growth rate,  
234  $\mu_{max}$ , and the maximum optical density,  $OD_{max}$ , were determined for nineteen experimental  
235 points. For each test, the experimental parameter was set at one level and the other two at their  
236 different levels in the field of study, for all possible combinations. Table 2 presents the factors  
237 and levels used in the central composite rotatable design, and the experimental matrix is  
238 shown in Table 3.

239 Figure 3 presents the kinetic parameters  $\lambda$ ,  $\mu_{max}$ , and  $OD_{max}$  calculated by the Baranyi and  
240 Roberts model. The results of the matrix, shown in Table 3, were analyzed using the statistical  
241 program Minitab 17, running on Windows 7. The coefficients of a quadratic polynomial  
242 model were estimated by multiple regression using the least squares method. The main,  
243 quadratic, and interaction effects, and the  $P$ -value corresponding to the Fisher-Snedecor test  
244 are listed in Table 4. The significance of each coefficient shown in Table 4 was checked by its

245 *P*-value: if the *P*-value < 0.05, the coefficient is significant, if the *P*-value > 0.05, the  
 246 coefficient does not contribute to the change of the response studied and was removed from  
 247 the regression equation. Coefficients with positive signs contribute to increase the studied  
 248 response and coefficients with negative signs reduce it. The quality of the predicted model is  
 249 evaluated by the coefficient of determination and the analysis of variances makes it possible  
 250 to estimate whether the predicted model is significant and adequate (Goupy and Creighton,  
 251 2006).

252 The *P*-value for the three models (0.000 for  $\lambda$ , 0.012 for  $OD_{max}$ , and 0.000 for  $\mu_{max}$ ) indicate  
 253 that the quadratic regression equations are significant and adequate. Also, the *P*-value due to  
 254 the lack of fit is > 0.05 (Table 1, Run 0) which indicates that the models have no bias. This  
 255 result is confirmed by comparing the values of the responses calculated by the model with  
 256 those obtained experimentally (Figure 3). The  $R^2$  values > 94 % (Table 5) for  $\lambda$ ,  $OD_{max}$ , and  
 257  $\mu_{max}$ , are explained by the corresponding regression. The equations for the three responses in  
 258 uncoded units (after eliminating the insignificant effects and recalculating the new  
 259 coefficients) are given by:

260

$$\begin{aligned} 261 \quad \lambda = & 4763 - 1116 \text{ pH} - 839.6 [(NH_4)_2SO_4]_0 - 4.893 [phenol]_0 + 59.82 [pH]^2 + 99.88 \\ 262 \quad & [(NH_4)_2SO_4]_0^2 - 0.01085 [phenol]_0^2 + 90.98 \text{ pH} [(NH_4)_2SO_4]_0 + 1.192 \text{ pH} [phenol]_0 \\ 263 \quad & \text{(Eq. 4)} \end{aligned}$$

$$\begin{aligned} 264 \quad \mu_{max} = & -78.98 + 14.1 \text{ pH} - 12.47 [(NH_4)_2SO_4]_0 + 0.5819 [phenol]_0 + 0.000281 [phenol]_0^2 - \\ 265 \quad & 0.11019 \text{ pH} [phenol]_0 + 0.10083 [(NH_4)_2SO_4]_0 [phenol]_0 \\ 266 \quad & \text{(Eq. 5)} \end{aligned}$$

$$\begin{aligned} 267 \quad OD_{max} = & 3.34 - 0.925 \text{ pH} - 0.0866 [(NH_4)_2SO_4]_0 + 0.001213 [phenol]_0 + 0.0657 [pH]^2 + \\ 268 \quad & 0.0553 [(NH_4)_2SO_4]_0^2 \\ 269 \quad & \text{(Eq. 6)} \end{aligned}$$

270

271 Microbial kinetics depend on several factors, including those chosen for the design of the  
272 experiment. In Eq. 4, pH has a negative effect on the response, so any increase in this  
273 parameter reduces the latency time. Similar results were found by Buchanan and Klawitter  
274 (1992) who studied the effect of the initial pH on the lag phase duration of the growth of  
275 bacteria *E. coli* 0157:H7 in the presence of NaCl. In this study, there is a different Na-  
276 containing compound present,  $\text{NaH}_2\text{PO}_4$ , but nevertheless, the bacteria seem to be better  
277 suited to an alkaline or neutral environment compared to an acidic environment. This effect is  
278 well illustrated in Figure 4.

279 The increase in the initial concentration of phenol is favorable within a certain limit, it is  
280 generally known that biodegradation by microbial cells is inhibited by phenol itself,  
281 particularly at higher concentrations, and the latency period increases with the initial  
282 concentration of phenol (Goupy and Creighton, 2006; Hamitouche and Bendjama, 2012)  
283 (Figure 4). It can be seen in Figure 4 that the initial phenol content has a negative effect and  
284 delays the maximum specific growth rate. Nevertheless, the highest  $OD_{\max}$  value is obtained  
285 for the higher initial phenol concentration.

286 The effect of the interactions is relative to the combined influence of the two parameters in  
287 question. In Figure 5, it is observed that even if the initial concentration of  $(\text{NH}_4)_2\text{SO}_4$  has no  
288 effect on the latency period, its interaction with the effect of pH has a positive effect on the  
289 response. Also, the pH and initial concentration of phenol interactions have a positive effect  
290 on the latency period.

291 In addition, the pH has no significant effect on the maximum specific growth rate in the  
292 studied domain, but its interaction with the initial concentration of phenol has a negative  
293 effect. In contrast, the interaction  $[(\text{NH}_4)_2\text{SO}_4]_0 * [\text{phenol}]_0$  has a positive effect on  $\mu_{\max}$ , so to

294 optimize this response, it is necessary to vary these last two factors in the same direction.

295 These results are represented in Figure 5.

296 The equations 4, 5, and 6 were used to draw the contour plots to define the effects of the  
 297 different factors on  $\lambda$ ,  $OD_{\max}$ , and  $\mu_{\max}$ . The contour plots in Figure 6 are useful for  
 298 determining desirable response values and operating conditions. The contour areas represent  
 299 constant responses, which correspond to  $\mu_{\max}$  varying from 0.1 up to 10 ( $\text{h}^{-1}$ ),  $OD_{\max}$  between  
 300 0.1 to 0.5, and  $\lambda$  going up to 200 h. The contours with the darkest green color indicate the  
 301 zone where response is the highest, while those with the darkest blue correspond to the  
 302 weakest responses.

303 It is remarkable that  $\mu_{\max}$  is dependent on the  $\text{pH}^*[(\text{NH}_4)_2\text{SO}_4]_0$  interaction, the highest value  
 304 is obtained in the areas corresponding to the upper left corner of the chart. This plot suggests  
 305 that it is possible to maximize  $OD_{\max}$  and  $\mu_{\max}$  or minimize  $\lambda$  under certain values of pH,  
 306 initial nitrogen, and phenol concentrations. The contours suggest that the response surface is a  
 307 rising ridge. Note that whatever the nitrogen concentration, these contours are identical, *i.e.*,  
 308 the nitrogen concentration does not influence this response.

309 The interaction  $\text{pH}^*[\text{phenol}]_0$  is not significant for  $OD_{\max}$  but highly significant for  $\lambda$  and  
 310  $\mu_{\max}$ . The highest growth rates are achieved at  $\text{pH} = 6$  for the solutions with the highest initial  
 311 phenol concentration and at  $\text{pH} = 8$  for solutions with lower initial phenol concentrations.

312 In Figure 6, the darkest green region is observed only in the upper right corner of the contour  
 313 curves of the  $OD_{\max}$  response when the specific growth rate is maximum for solutions with  
 314 the highest initial concentrations of phenol and ammonium sulfate. It is also noted on the  
 315 contour diagram that the smallest latency value ( $\lambda = 1$  h) is obtained for the following  
 316 conditions:

- 317 • For neutral pH, in the solution with 1 g/L of ammonium sulfate and 50 mg/L of phenol

318 Also, in the same logic, the optimal values for  $\mu_{\max}$  can be estimated for:

319 • pH = 6, in the solution with 2 g/L of ammonium sulfate and 200 mg/L of phenol  
320 The contour curves for  $OD_{\max}$  indicate that the optimum is obtained at pH = 6 for an initial  
321 phenol concentration equal to 200 mg/L and for an initial ammonium sulfate content equal to  
322 2 g/L, similar values to those corresponding to  $\mu_{\max}$ .

323 The optimal values for *E. coli* growth parameters are:  $\lambda_{\min} = 21.08$  h,  $\mu_{\max} = 8.68$  h<sup>-1</sup> and  
324  $OD_{\max} = 0.39$  at pH = 6.3 for  $[phenol]_0 = 200$  mg/L and  $[(NH_4)_2SO_4]_0 = 1.33$  g/L, values given  
325 by the Minitab Software 17.

326

### 327 3.4. Effect of initial phenol concentration

328 For the initial concentrations of phenol equal to 562 and 963 mg/L, bacterial growth and the  
329 residual concentration of phenol (Figure 7a) were monitored as a function of time. In  
330 agreement with previous works (Câmara et al., 2019; Hamitouche and Bendjama, 2012; Peng  
331 et al., 2018; Stoilova et al., 2006; Wang et al., 2019), the increase of the initial concentration  
332 of phenol automatically leads to the increase of the biomass produced at the end of the  
333 exponential phase.

334 The effect of the initial phenol concentration on the latency period is shown in Figure 7b. It is  
335 observed that the latency time increases with initial phenol concentration, which leads to the  
336 conclusion that the substrate has an inhibitory effect on the growth of bacteria (Ucun et al.,  
337 2010).

338 The results presented in this work are innovative. While most of the previous works first  
339 acclimatize the microorganisms to varying concentrations of phenol before carrying out the  
340 biodegradation tests, all the growth curves recorded in this work are obtained for cultures of  
341 bacteria taken directly without prior acclimatization. This is only the second work that deals  
342 with an *E. coli* strain that is able to biodegrade phenol without acclimatization, the other being  
343 Fayidh et al. (2015).

344

**345 4. Conclusions**

346 This study demonstrated the ability of *Escherichia coli* RV412\_A1\_2010\_06a LBK to remove  
347 different phenol concentrations up to 963 mg/L. Modeling using the Surface Response  
348 Methodology was performed to study the effects of pH and initial concentrations of  
349 ammonium sulfate and phenol on growth parameters (latency time,  $\lambda$  (h), maximum specific  
350 growth rate,  $\mu_{\max}$  ( $\text{h}^{-1}$ ), and maximum optical density,  $OD_{\max}$ ). Second order models were  
351 developed to predict growth parameters. Thanks to ANOVA software, it has been statistically  
352 demonstrated that approximately 98.8%, 99.6%, and 94.5% of the responses for  $\mu_{\max}$ ,  $\lambda$ , and  
353  $OD_{\max}$  respectively are explained by the regression equations of these models. The optimal  
354 values of growth parameters are  $\lambda_{\min} = 21.08$  h,  $\mu_{\max} = 8.68$   $\text{h}^{-1}$ , and  $OD_{\max} = 0.39$  at pH = 6.3  
355 for  $[\text{phenol}]_0 = 200$  mg/L and  $[(\text{NH}_4)_2\text{SO}_4]_0 = 1.33$  g/L. As a result of this work, this strain will  
356 be acclimatized to large amounts of phenolic pollutants and other more pathogenic pollutants  
357 such as pharmaceutical compounds in a future work.

358

**359 Acknowledgements:**

360 S. D. L. is grateful to F.R.S.-F.N.R.S. for her Senior Research Associate position. The authors  
361 also acknowledge the Ministère of Algeria for financial support.

362

**363 Compliance with ethical standards**

364 Conflict of interest: the authors declare that they have no conflicts of interest.

365

**366 Data availability**

367 The raw/processed data required to reproduce these findings cannot be shared at this time as  
368 the data also forms part of an ongoing study.

369

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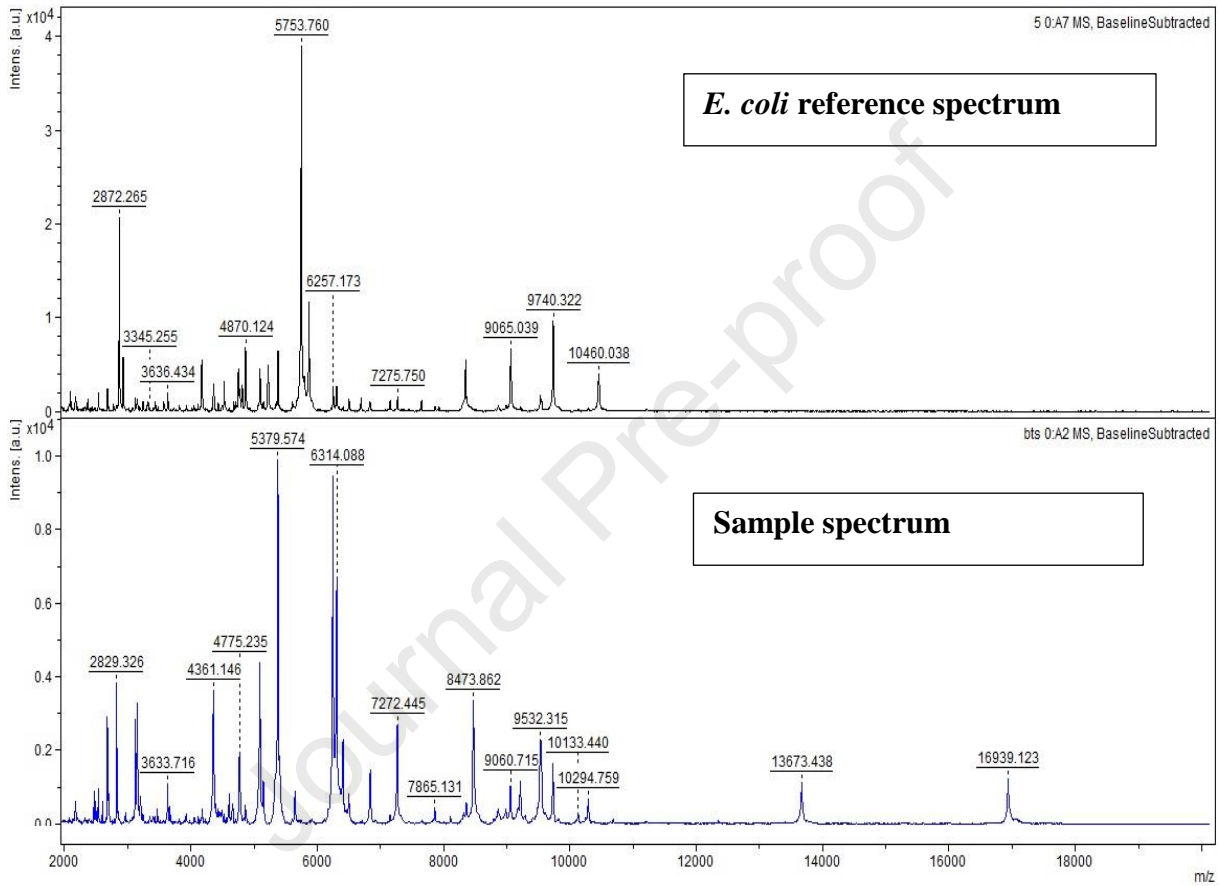
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542 **Figure 1:** Matrix assisted laser desorption ionization - time of flight (MALDI-TOF) spectrum543 obtained from study samples compared to *Escherichia coli* reference spectrum.

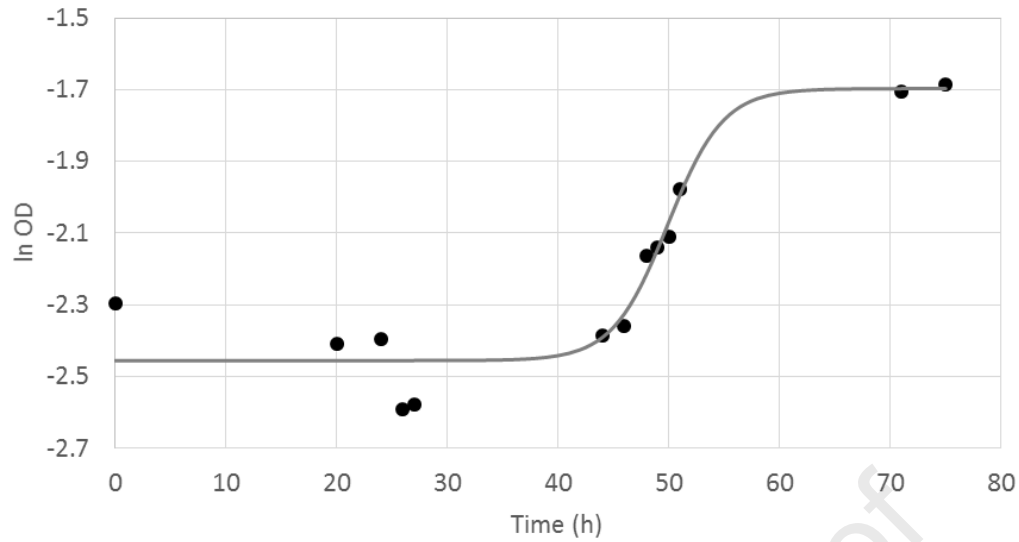
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550 **Figure 2:** Growth curve of *Escherichia coli* at 30°C fitted with the Baranyi and Roberts  
551 model (Imadalou-Idres et al., 2018; Romano et al., 2007), OD = optical density.

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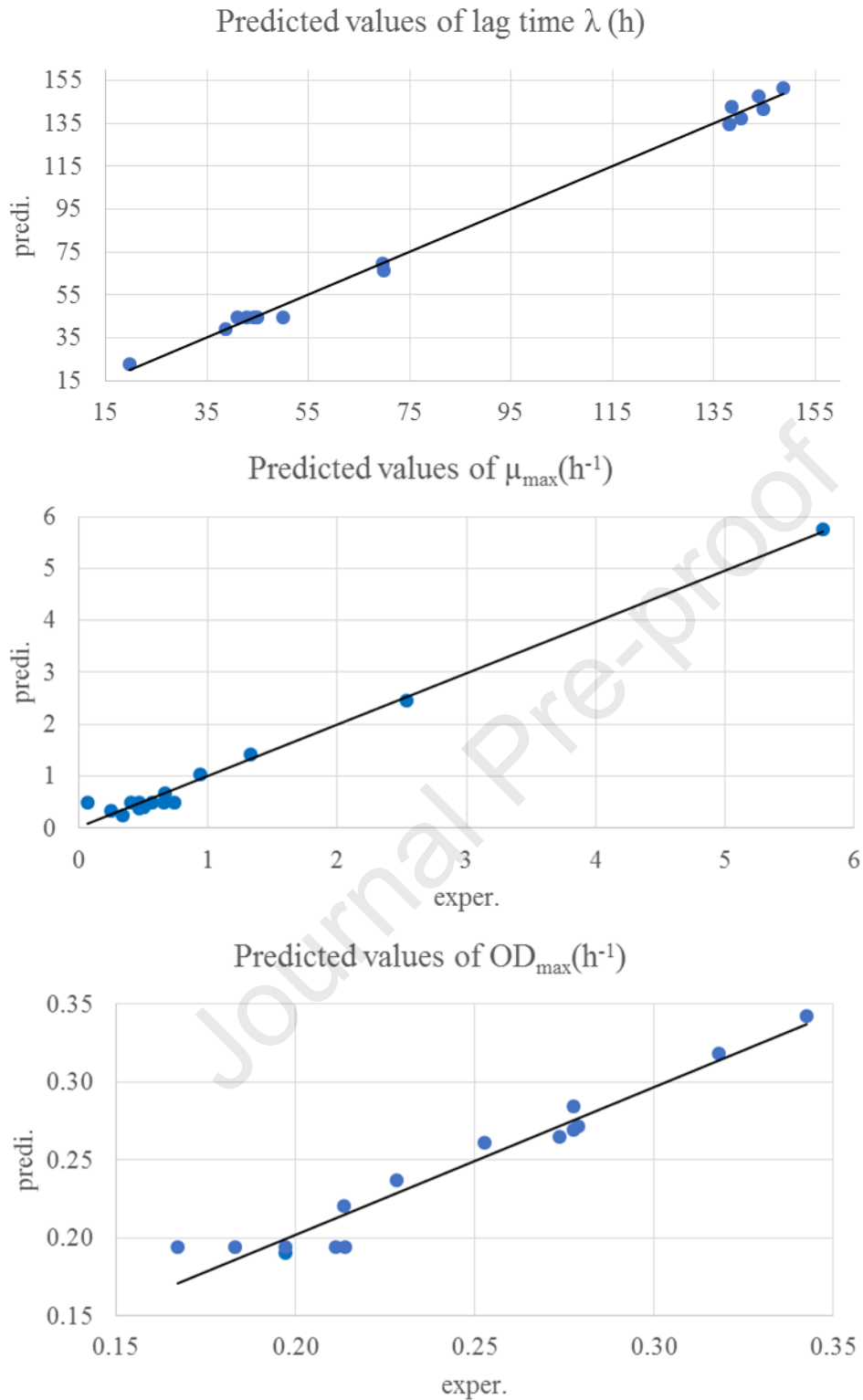
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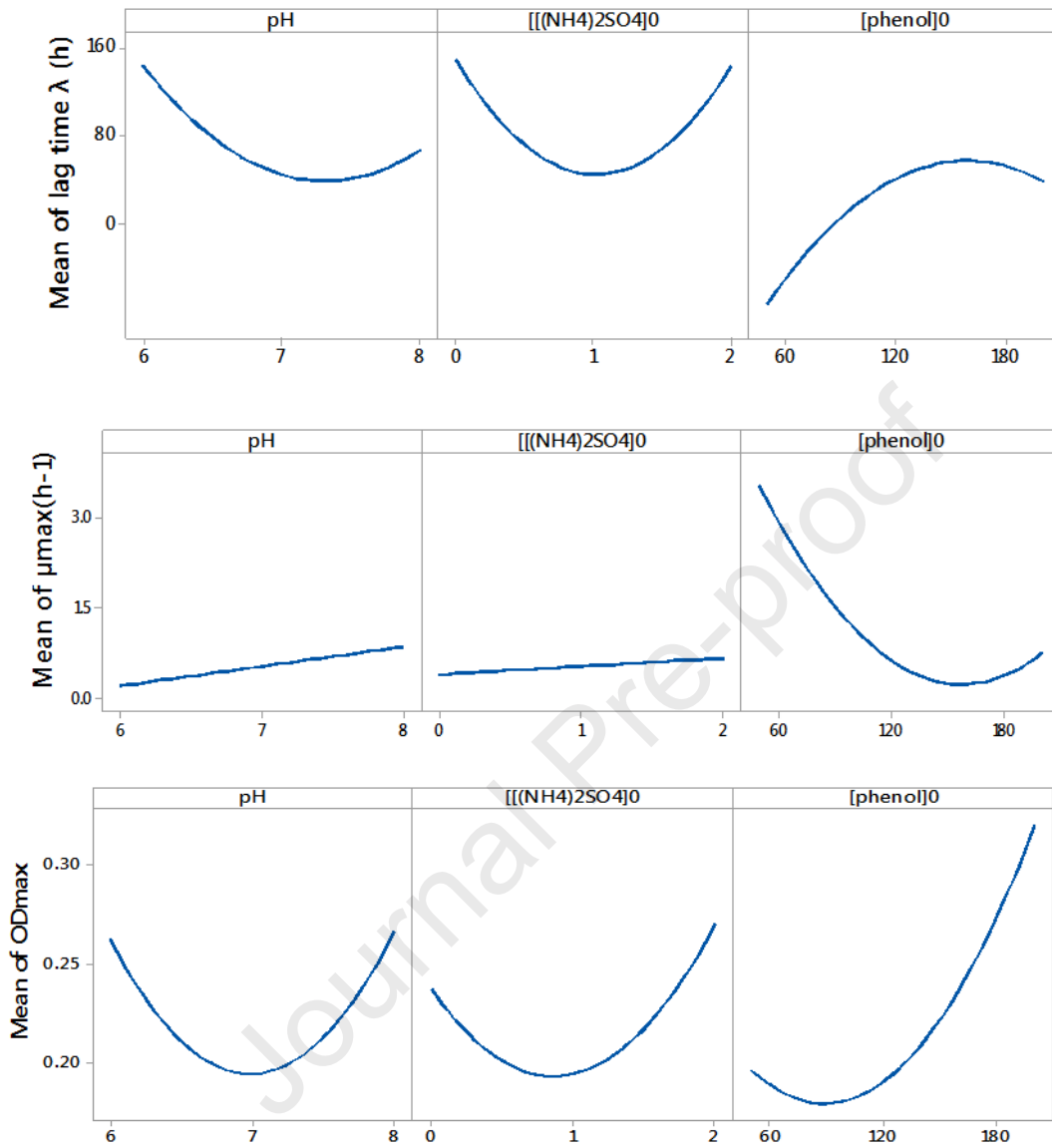
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563 **Figure 3:** Predicted values of responses using the Baranyi and Roberts model for latency564 period,  $\lambda$ , the maximum specific growth rate of *E. coli*,  $\mu_{\max}$ , and the maximum optical565 density,  $OD_{\max}$ .

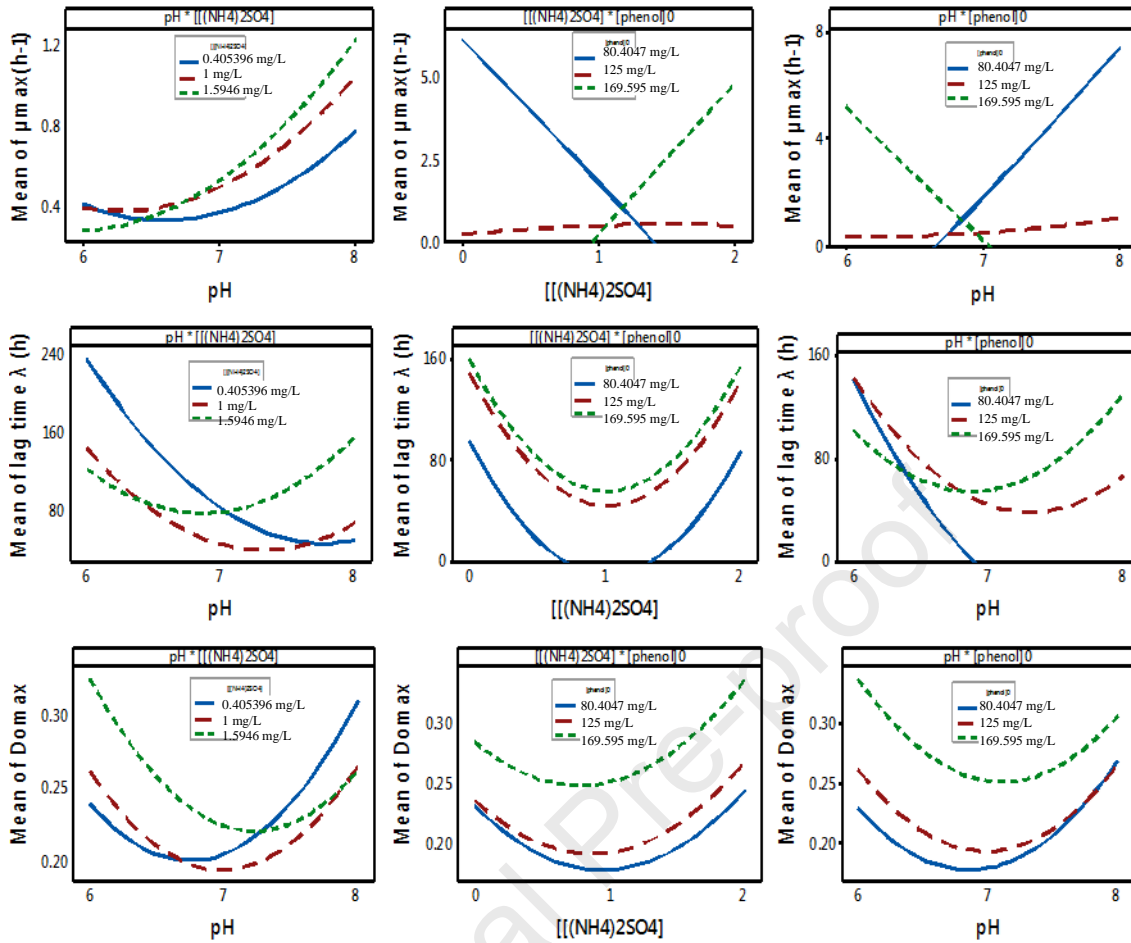
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568 **Figure 4:** Main effects plot of mean latency time,  $\lambda$ , maximum specific growth rate,  $\mu_{\max}$ , and  
 569 maximum optical density,  $OD_{\max}$ .

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573 **Figure 5:** Plots of interaction effects for latency time,  $\lambda$ , maximum specific growth rate,  $\mu_{\max}$ ,  
574 and maximum optical density,  $OD_{\max}$ .

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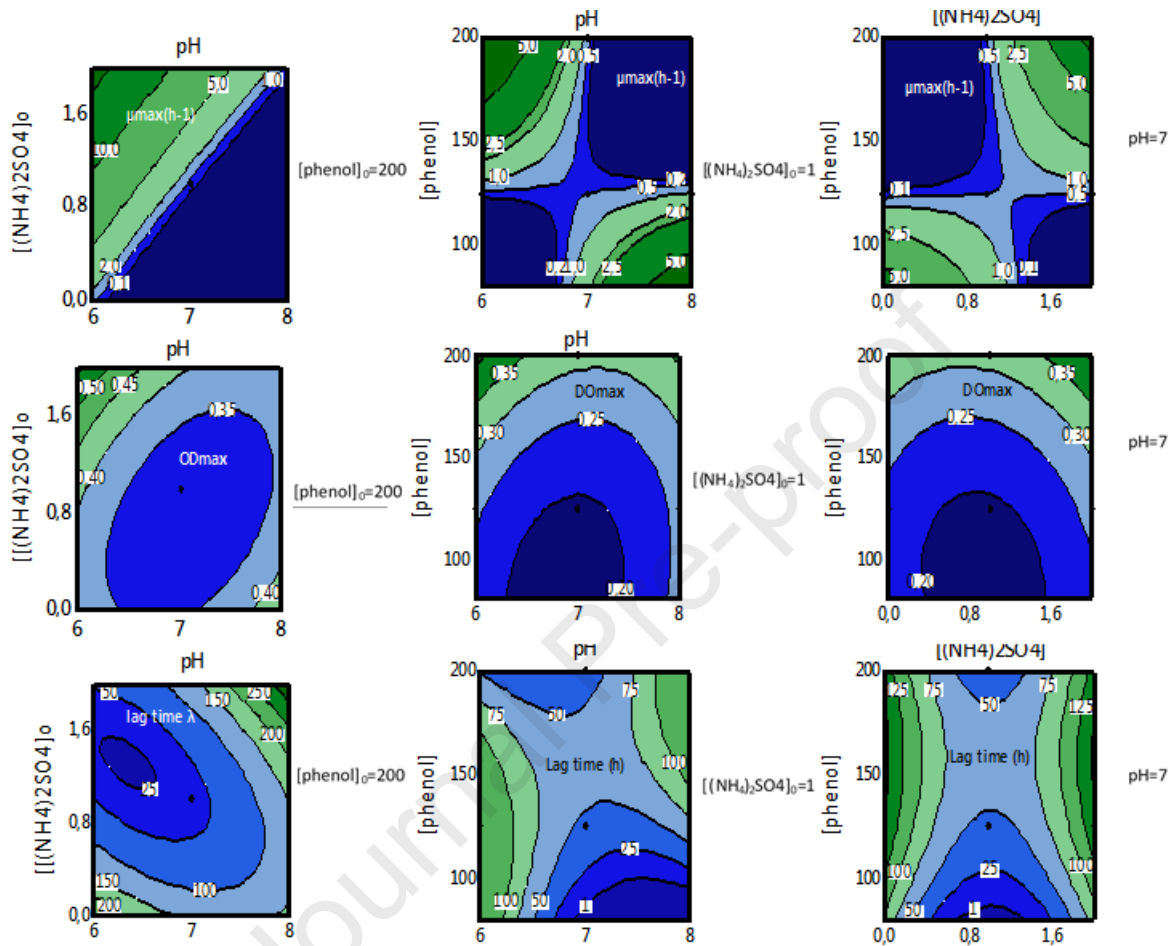
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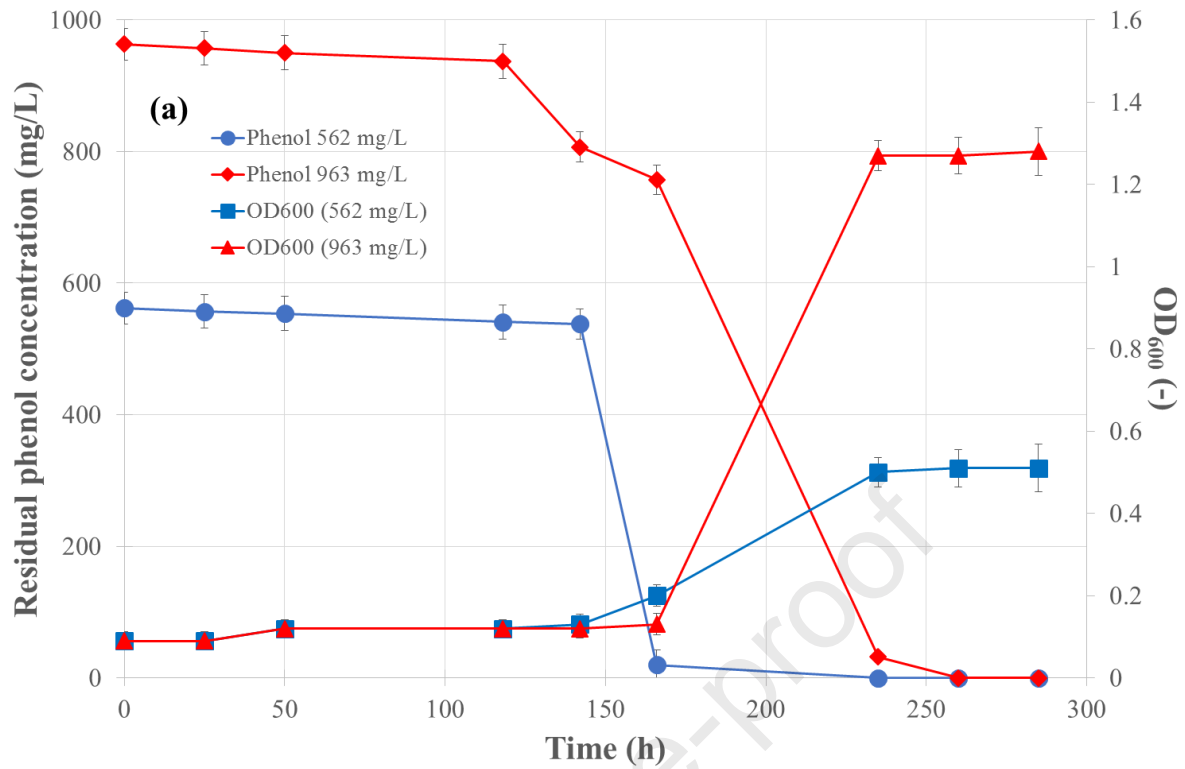


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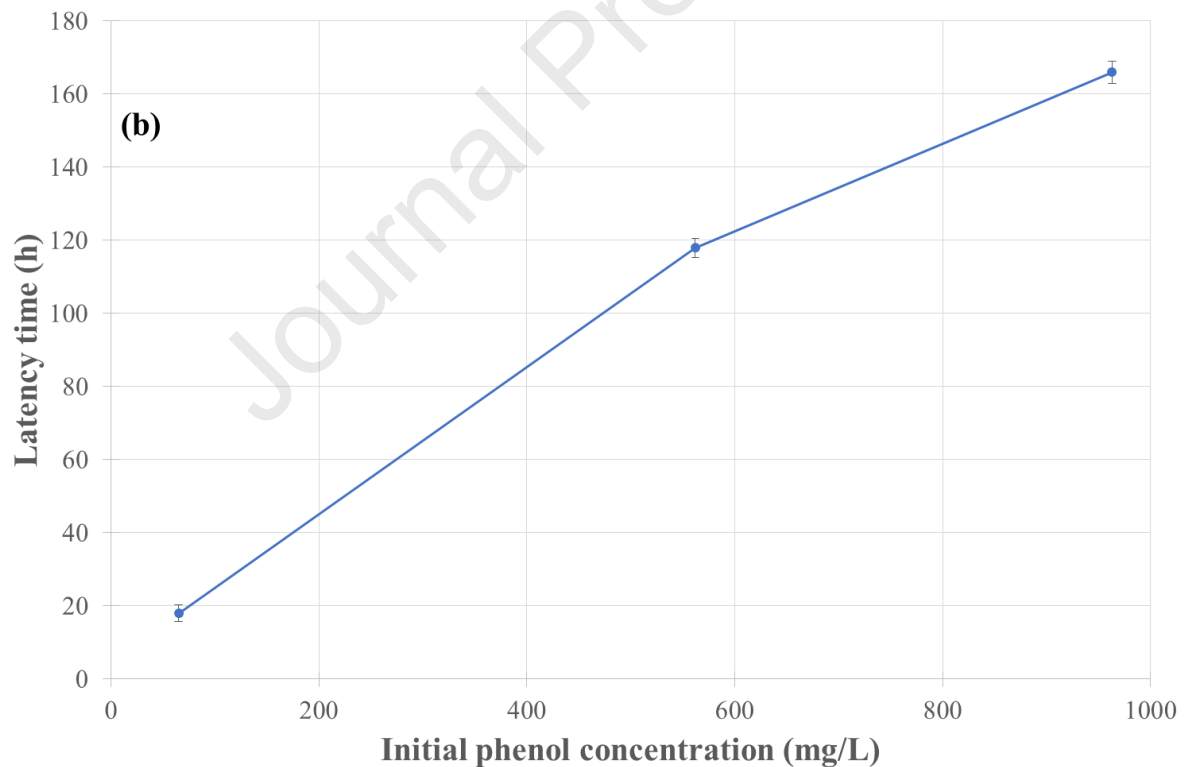
589 **Figure 6:** Contour plots for the three responses showing latency time,  $\lambda$ , maximum specific590 growth rate,  $\mu_{\max}$ , and maximum optical density,  $OD_{\max}$ .

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595 **Figure 7:** (a) Residual concentration of phenol and optical density as a measurement of  
 596 bacterial growth as a function of time. (b) Effect of the initial phenol concentration on the  
 597 latency time.

Journal Pre-proof

599 **Table 1:** Analysis of variances for the Gompertz and Baranyi and Roberts primary models.

Model		Gompertz	Baranyi and Roberts
Run 1: pH = 6.4, $[(NH_4)_2SO_4]_0 = 0.405$ , $[phenol]_0 = 80.4$	RMSE	0.0799	0.0526
	R <sup>2</sup> adj	0.9257	0.9678
Run 4: pH = 7.6, $[(NH_4)_2SO_4]_0 = 1.59$ , $[phenol]_0 = 80.4$	RMSE	0.0527	0.0526
	R <sup>2</sup> adj	0.9747	0.9748
Run 5: pH = 6.4, $[(NH_4)_2SO_4]_0 = 0.405$ , $[phenol]_0 = 170$	RMSE	0.0671	0.0681
	R <sup>2</sup> adj	0.9454	0.9438
Run 7: pH = 6.4, $[(NH_4)_2SO_4]_0 = 1.59$ , $[phenol]_0 = 170$	RMSE	0.1524	0.1524
	R <sup>2</sup> adj	0.9562	0.9562
Run 8: pH = 7.6, $[(NH_4)_2SO_4]_0 = 1.59$ , $[phenol]_0 = 170$	RMSE	0.0667	0.0672
	R <sup>2</sup> adj	0.9528	0.9520
Run 9: pH = 6.0, $[(NH_4)_2SO_4]_0 = 1.00$ , $[phenol]_0 = 125$	RMSE	0.0945	0.0924
	R <sup>2</sup> adj	0.9329	0.9359
Run10: pH = 8.0, $[(NH_4)_2SO_4]_0 = 1.00$ , $[phenol]_0 = 125$	RMSE	0.1021	0.1036
	R <sup>2</sup> adj	0.9556	0.9544
Run11: pH = 7.0, $[(NH_4)_2SO_4]_0 = 0$ , $[phenol]_0 = 125$	RMSE	0.0747	0.0742
	R <sup>2</sup> adj	0.9320	0.9328
Run12: pH = 7.0, $[(NH_4)_2SO_4]_0 = 2.00$ , $[phenol]_0 = 125$	RMSE	0.0734	0.0780
	R <sup>2</sup> adj	0.9284	0.9193
Run 14: pH = 7.0, $[(NH_4)_2SO_4]_0 = 1$ , $[phenol]_0 = 200$	RMSE	0.0785	0.0786
	R <sup>2</sup> adj	0.9857	0.9856
Run 0: pH = 8, $[(NH_4)_2SO_4]_0 = 1$ , $[phenol]_0 = 125$	RMSE	0.2222	0.2218
	R <sup>2</sup> adj	0.6616	0.6629
	Lack of fit (P-value)	0.6360	0.6570

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603 **Table 2:** Experimental range for a central composite rotatable design

Parameters	Coded values				
	-1.68	-1	0	1	1.68
X1 pH	6.0	6.4	7.0	7.6	8.0
X2 $[(NH_4)_2SO_4]_0$ , g/L	0	0.4	1.0	1.6	2.0
X3 $[phenol]_0$ , mg/L	50.0	80.4	125	170	200

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605 **Table 3:** Matrix of experiments with coded and non-coded factors

Run	Coded parameters			Non coded parameters		
	X1	X2	X3	pH	$[(NH_4)_2SO_4]_0$	$[phenol]_0$
1	-1	-1	-1	6.4	0.4	80.4
2	1	-1	-1	7.6	0.4	80.4
3	-1	1	-1	6.4	1.6	80.4
4	1	1	-1	7.6	1.6	80.4
5	-1	-1	1	6.4	0.4	170.0
6	1	-1	1	7.6	0.4	170.0
7	-1	1	1	6.4	1.6	170.0
8	1	1	1	7.6	1.6	170.0
9	-1.68	0	0	6.0	1.0	125.0
10	1.68	0	0	8.0	1.0	125.0
11	0	-1.68	0	7.0	0.0	125.0
12	0	1.68	0	7.0	2.0	125.0
13	0	0	-1.68	7.0	1.0	50.0
14	0	0	1.68	7.0	1.0	200.0
15	0	0	0	7.0	1.0	125.0
16	0	0	0	7.0	1.0	125.0
17	0	0	0	7.0	1.0	125.0
18	0	0	0	7.0	1.0	125.0
19	0	0	0	7.0	1.0	125.0

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**608 Table 4:** Estimated Coefficients (coded units) and their *P*-values

Parameters Term	$\mu_{\max}$ ( $\text{h}^{-1}$ )		$\lambda$ (h)		$OD_{\max}$	
	Coef	<i>P</i> -Value	Coef	<i>P</i> -Value	Coef	<i>P</i> -Value
Constant	0.501	0.008	44.55	0.000	0.19456	0.000
pH	0.1912	<u>0.094<sup>a</sup></u>	-22.64	0.000	0.00117	<u>0.876<sup>a</sup></u>
$[(\text{NH}_4)_2\text{SO}_4]_0$	0.0793	<u>0.432<sup>a</sup></u>	-1.82	<u>0.304<sup>a</sup></u>	0.00961	<u>0.234<sup>a</sup></u>
$[\text{phenol}]_0$	-0.787	0.002	32.88	0.000	0.0364	0.015
$[\text{pH}]^2$	0.0775	<u>0.364<sup>a</sup></u>	21.15	0.000	0.02427	0.01
$[(\text{NH}_4)_2\text{SO}_4]_0^2$	-0.046	<u>0.58<sup>a</sup></u>	35.31	0.000	0.02063	0.018
$[\text{phenol}]_0^2$	0.528	0.009	-21.57	0.000	0.02206	<u>0.075<sup>a</sup></u>
$\text{pH} * [(\text{NH}_4)_2\text{SO}_4]_0$	0.086	<u>0.688<sup>a</sup></u>	32.16	0.000	-0.0197	<u>0.257<sup>a</sup></u>
$\text{pH} * [\text{phenol}]_0$	-2.967	0.000	31.6	0.000	-0.0104	<u>0.568<sup>a</sup></u>
$[(\text{NH}_4)_2\text{SO}_4]_0 * [\text{phenol}]_0$	2.718	0.000	0.44	<u>0.924<sup>a</sup></u>	0.006	<u>0.74<sup>a</sup></u>

**609** <sup>a</sup> *P*-values (underlined) indicate that the corresponding coefficient is not significant.

**610**

**611 Table 5:** Analysis of variance for the secondary models (the first three responses are based on  
**612** the Baranyi and Roberts primary model)

Parameters →	$\mu_{\max}$ ( $\text{h}^{-1}$ )	$\lambda$ (h)	$OD_{\max}$ (-)
Source of variation		<i>P</i> -Value	
Model	0.000	0.000	0.012
Linear	0.007	0.000	<u>0.054<sup>a</sup></u>
Square	0.037	0.000	0.024
2-Way Interaction	0.000	0.000	<u>0.312<sup>a</sup></u>
Lack-of-Fit	<u>0.375<sup>a</sup></u>	<u>0.113<sup>a</sup></u>	<u>0.325<sup>a</sup></u>
$R^2$ (%)	98.8	99.6	94.5

**613** <sup>a</sup> *P*-values (underlined) indicate that the corresponding coefficient is not significant.

**614**

**615**

- Biodegradation process of phenol in a batch reactor using *Escherichia coli*.
- Optimization of the growth parameters thanks to Baranyi and Roberts' equations
- Optimization of experimental conditions using secondary models based on a central composite rotatable design
- Degradation up to 963 mg/L of phenol in 250 h without acclimatization of *Escherichia Coli*.

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Conflict of interest: The authors declare that they have no conflicts of interest.

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