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Estimating the Efficacy of Mild Heating Processes taking into Account Microbial Non-linearities: a Case Study on the Thermisation of a Food Simulant

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Estimating the efficacy of mild heating processes taking into account microbial non-linearities: a case study on the thermisation of a food simulant

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2 linearities: a case study on the thermisation of a food simulant

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28 Abstract

58	Keywords: Thermisation, Listeria innocua, non-linearity, modelling
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41 42	account the observed non-intearity.
40	rmany, the equivalent nothing times of different temperatures are calculated taking into
39 40	Finally, the equivalent holding times of different temperatures are coloulated taking inte
38 20	et al., (2000) model. Thereafter, dynamic temperature profiles (targeting at 54 and 5/ $^{\circ}$ C)
51 29	isothermal conditions of 58, 60, 63 and 66°C by the use of the log-linear and the Geeraerd (2000) model. Thereafter, demension to (1000) for (1000) model.
36 27	<i>unnocua</i> CLIP 20-595 are used for estimating the inactivation parameters under $1 + 1 + 1 + 1 + 1 + 1 + 1 + 1 + 1 + 1 $
35	kinetics. An illustrative example is given in which microbial inactivation data of <i>Listeria</i>
34	constant, z, for log-linear or specific types of non log-linear microbial inactivation
33	a predefined microbial reduction at a reference temperature and a known thermal resistant
32	to estimate the efficiency of a mild heat process. MH-value is the time needed to achieve
31	commonly used sterilisation, pasteurisation and cook value (F, P, C-value), is calculated
30	compared in this work. The Mild Heat value (MH-value), an alternative approach to the
29	Traditional and novel approaches for the calculation of the heat treatment efficiency are

59 Introduction

60 When microbial inactivation processes are described as log-linear, a linear relationship 61 between the logarithm of the microbial population level (in absolute value or relative to 62 the initial value) and the treatment time are considered. If log-linearity is indeed being 63 observed, the thermal death time, F-value is used as a basis for comparing heat 64 sterilisation procedures. The F-value (Eq. (1)) (Ball, 1923) is defined as the time required 65 in order to achieve a specific reduction in microbial numbers at a given temperature and it 66 thus represents the total time-temperature combination received by a food. Similarly, the 67 *P*-value and the *C*-value are the corresponding thermal death value under pasteurisation 68 and cooking conditions, respectively (Pittia, Furlanetto, Maifreni, Mangina & Rosa, 69 2008).

$$F_{T_{ref}} = \int_{0}^{t} 10^{(T(t) - T_{ref})/z} dt \quad (1)$$

70

71 Eq. (1), as mentioned above, is valid if the survival curve obeys first order kinetics. 72 Despite the world-wide use of this approach especially in the canning industry for the so-73 called '12D process' of the proteolytic strains (Group I) of *Clostridium botulinum* spores 74 (Stumbo, 1965; ICMSF, 1996), a lot of deviations from log-linearity have been observed 75 (e.g., Corradini, Normand and Peleg (2005)). As other authors acknowledge, the success 76 of the canning industry in using the *F*-value as a measure of the heat processes efficacy 77 could be attributed to over-processing and not to the calculation method's correctness 78 (Corradini, Normand & Peleg, 2006). These deviations are evident particularly at lower 79 temperatures than the sterilisation ones, and for vegetative cells (Valdramidis, Geeraerd, 80 Bernaerts & Van Impe, 2006; Huang, 2009; Miller, Gil, Brandao, Teixeira & Silva, 81 2009).

A mild heat treatment often applied in the dairy industry is thermisation of milk destinated for cheese making. This process has milder effects on the raw milk flora and the functionality of milk caseins and salts than pasterurisation (Samelis et al., 2009). Thermisation is applied at temperatures that range between 52 to 67°C for a treatment time of few seconds, i.e., 20 s, to about half an hour (Zehetner, Bareuther, Henle & Klostermeyer, 1996; McKellar & Piyasena, 2000; Christiansen, Nielsen, Vogensen,
Brogren & Ardo, 2006; Levieux, Geneix & Levieux, 2007; Samelis et al., 2009) while
according to the council directive 92/46/EEC of 16 June 1992, thermisation is the heating
of raw milk for at least 15 s at a temperature between 57°C and 68°C such that after
treatment the milk shows a positive reaction to the phosphate test.

92 When focusing on microbiological safety, the accurate description of the kinetics of the 93 target pathogenic microorganism (or a surrogate of a pathogenic target microorganism) is 94 essential. Milk designed for milk based products should receive a heat treatment process to reduce the probability of survival of L. monocytogenes by at least a factor of 10^4 (FIL-95 96 IDF, 1994). Taking a safety margin into account, usually a log reduction of 6 is 97 considered (Claeys, Van Loey & Hendrickx, 2002). Estimation of the time to achieve this 98 log reduction is a very critical issue for the design and application of an efficient heat 99 treatment.

100 Over the last 30 years a number of inactivation models have been developed aiming at 101 describing non-log-linear microbial inactivation kinetics. An overview of inactivation 102 models portraying eight common type curves is given by Geeraerd, Valdramidis and Van 103 Impe, (2005). The development of these models raises the need of redefining the thermal 104 death time by a modelling approach that includes the possibility of non-log-linear 105 microbial survival curves, especially in cases of mild heat treatments, like thermisation. 106 Recent studies in the broader field of heat processing suggested alternative approaches 107 for evaluating the efficacy of a process when assuming that microbial heat resistances 108 follow a weibulian frequency distribution model (Mafart, Couvert, Gaillard & Leguerinel, 109 2002; Corradini et al., 2006; Sant'Ana, Rosenthal & Massaguer, 2009). Nevertheless, 110 these approaches take into account only two types of non-log-linearity (i.e., concave, 111 convex) and are not retaining classical parameters (like the z - value) for evaluating the 112 achieved microbial reduction.

113 The main objective of this study was to calculate a value for the efficacy of a mild 114 heating process similar to thermisation in which inactivation kinetics are not log-linear 115 and to test the approach for the mild heat treatment of thermisation for a (model) liquid 116 food system. *L. innocua* is considered as the surrogate safety target attributed for the 117 studied simulant liquid food. The final objective was to develop an alternative (to the classical *F*, *P*, *C* values) mathematical expression in order to evaluate the efficiency of a
thermal treatment if additional environmental or physiological factors are considered.

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- 121
- 122 Materials and methods
- 123
- 124 Modelling approaches

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129

In this study a general expression of the microbial inactivation kinetics (Van Impe,
Poschet, Geeraerd & Vereecken, 2005), was considered. This expression is described as
follows.

$$\frac{dN}{dt} = -k(N, \langle env \rangle, \langle phys \rangle) \cdot N \quad (2)$$

N is the cell density of the microbial species (cfu/mL), $\langle env \rangle$ denotes the actual (micro)environmental conditions (not or only slightly influenced by the microbial evolution) such as temperature, high pressure, salt concentration, water activity, etc and $\langle phys \rangle$ is the physiological state of the species, for instance, as influenced by the temperature history. This expression can then be coupled with differential equations that describe the dynamics of the physiological state parameters e.g., Geeraerd, Herremans & Van Impe, (2000).

A sound set of differential equations, which is a sub-case of Eq. (2), and describes the microbial inactivation kinetics by incorporating physiological adjustments during the microbial inactivation experiments is the dynamic, non-log-linear model of (Geeraerd et al., 2000). This model is constructed for microbial inactivation by mild heating.

$$\frac{dN}{dt} = -k_{\max} \cdot \left(\frac{1}{1+C_c}\right) \cdot \left(1-\frac{N_{res}}{N}\right) \cdot N \quad (3)$$

$$\frac{dC_c}{dt} = -k_{\max} \cdot C_c \quad (4)$$

143 Herein, N represents the microbial cell density [cfu/mL], C_c is related to the physiological 144 state of cells [-], k_{max} denotes the specific inactivation rate [1/min] and N_{res} the residual 145 population density [cfu/mL]. This is a model for describing non-linearities that 146 incorporate shoulder and/or tailing effects and it automatically reduces to log-linear 147 inactivation kinetics if the data do not include these effects. Although there are more 148 known survivor curve shapes for vegetative bacterial cells (Geeraerd et al., 2005) the 149 current modelling approach is built based on the features of the microbial data of the case 150 study presented hereunder.

Thermal inactivation parameters, i.e., the asymptotic decimal reduction time (*AsymD*value) (Juneja, Eblen & Marks, 2001) and the thermal resistance constant (*z*-value), i.e., the temperature change required to achieve a tenfold change in *AsymD*-value, integrated into the Bigelow model (Eq. (5)) yields predictions for the specific inactivation rate at a given temperature.

156

$$k_{\max}(T) = \frac{\ln 10}{AsymD_{ref}} \cdot \exp\left(\frac{\ln 10}{z} \cdot (T - T_{ref})\right) \quad (5)$$

157

Herein, $AsymD_{ref}$ [min] is the asymptotic decimal reduction time at the reference temperature T_{ref} [°C]. Observe that D_{ref} of the original Bigelow equation is replaced by $AsymD_{ref}$ as it describes the negative inverse of the slope of the log linear part of the inactivation curve.

- 163 Eqs (2)-(5) are the main set of equations for calculating the mild heat pasteurisation
 164 value, *MH*-value (see Results).
- 165 The explicit version of Eqs (3)-(5) is as follows (Eq. (5) is inserted in the explicit version
- 166 of Eqs (3), (4)).
- 167

$$\log(N(t)) = \log((10^{\log(N(0))} - 10^{\log(Nres)}) \cdot \exp\left(-\frac{\ln 10}{AsymD_{ref}} \cdot \exp\left(\frac{\ln 10}{z} \cdot (T - T_{ref})\right) \cdot t\right)$$
$$\cdot \frac{1 + 10^{\log(C_c(0))}}{1 + 10^{\log(C_c(0))} \cdot \exp\left(-\frac{\ln 10}{AsymD_{ref}} \cdot \exp\left(\frac{\ln 10}{z} \cdot (T - T_{ref})\right) \cdot t\right)} + 10^{\log(Nres)}\right) \quad (6)$$

169 In order to demonstrate the validity of the modelling approach that evaluates the *MH*-170 value some illustrative examples are given. Firstly, a dynamic profile representing 171 thermisation was generated and secondly, the static temperatures of different time 172 treatments for achieving the same microbial reduction were evaluated. Both studies were 173 performed for a given microbial inactivation kinetics (see hereunder).

174

175 *Case study*

176 Microbial essay

177

178 The studied methodological approach is illustrated based on data of microbial kinetics of 179 Listeria innocua CLIP 20-595 originating from Peroval, Portanguen and Kondjoyan, 180 (2004). Summarising, the heat resistance of *Listeria innocua* was studied by the use of 181 100 µL of the cell suspension sealed in sterile glass capillary tubes. The tubes were 182 immersed in a thermostat controlled circulating water bath at temperatures of 58, 60, 63, 183 66, 68 and 70°C. Come-up times, which were in any case very small, were included as 184 part of the total heating time used to calculate the inactivation parameters. Decimal serial 185 dilutions of the samples were made in TS medium and surface plated in duplicate on 186 PALCAM agar (Merck). Plates were incubated for 24 -48h at 37°C and colony-forming

187 units (cfu) were enumerated. Heat resistance experiments were carried out in triplicate.

188

189 Parameter estimation

(i)

(ii)

190

191 The microbial inactivation parameters under the isothermal conditions were estimated:

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- 193 194

by assuming first order inactivation kinetics for all the inactivation data (Eq. (3) when omitting the second and third factor of the right hand side)

by taking into account the appearing shoulder and tailing effects (Eqs (3), (4)),

(iii) by assuming first order inactivation kinetics only for the log-linear portion ofthe inactivation kinetics.

197 The last two approaches are the classical approaches for parameter estimation, but in 198 case (iii) there is some subjectivity on the choice of the microbial data that belong to 199 the log-linear portion. In this study the choice was made taking into account the 200 information coming from case (i) (i.e., initiation of the log-linear portion at the end of 201 the shoulder up to the beginning of the tailing). Consequently, the parameters of the 202 last procedure will be similar to the parameters of the log-linear portion of case (i). 203 Hereupon, these three different case studies have been used in order to evaluate the 204 equivalent isothermal temperature of a dynamic temperature profile.

205

206 In order to identify the model parameters a so called global identification making use of 207 all static experiments in one step was implemented. So for example in the case (i) Asym D_{ref} , z, log $C_c(0)$, and logN(0) (one for each temperature), log N_{res} (one for each 208 209 temperature) (Eq. (6)) were estimated. The selected reference temperature was chosen to be equal to 62° C, as the optimal choice to minimize the uncertainty on Asym D_{ref} was in 210 211 the middle of the studied temperature range as commonly chosen (Poschet, Geeraerd, 212 Van Loey, Hendrickx & Van Impe, 2005). Parameters were estimated based on the 213 minimisation of the Sum of Squared Errors (SSE).

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- 215

216 *Thermisation temperature profile*

218 Thermisation experiments were conducted in a pilot scale retort following the same 219 procedure described in (Patras, Tiwari, Brunton & Butler, 2009). The prepared cans (75 × 220 110 mm, WEI/WEISS03, Germany) were filled with the same suspension used for the 221 microbial studies, i.e., Tryptone-Salt (TS) medium 0.1% w/v, 0.85% w/v NaCl) and were 222 loaded into the pilot scale retort (Barriguand Steriflow, Roanne, France). Sample core 223 temperature profiles were recorded during the process, using an Ellab E-Val TM TM9608 224 data module (Ellab [UK] Ltd., Norfolk, England) connected to a laptop. A standard Ellab 225 SSA-12080-G700-TS temperature probe was inserted through an Ellab GKM-13009-226 C020 packing gland (20 mm) into the a can to record the temperature cycle. Temperature 227 was monitored every 10 s. The samples were heated targeting at a final temperature of 54 and 57°C. Prior to any canning experiment, all Ellab unit probes were calibrated against a 228 229 JOFRA (ATC-155B) calibration unit.

230

231 Results & Discussion

232 *Defining the MH-value* 233

MH-value was defined as the time needed to achieve a predefined microbial reduction at a reference temperature and a known thermal resistant constant, *z*, when microbial inactivation kinetics is not linear. Mathematically, this expression can be calculated by deriving the achieved microbial reduction at given temperature conditions. So if the nonlog-linear microbial kinetics are described by the (Geeraerd et al., 2000) Eqs (3)-(5) then the achieved microbial reduction is given as follows.

$$\int_{0}^{t} \frac{dN(t)}{N(t)} = -\int_{0}^{t} k_{\max} \cdot \left(\frac{1}{1+C_{c}}\right) \cdot \left(1-\frac{N_{res}}{N}\right) \cdot dt$$

240

$$\ln\left(\frac{N(t)-N_{res}}{N(0)-N_{res}}\right) = -\frac{\ln 10}{\operatorname{Asym} D_{ref}} \cdot \int_0^t \exp\left(\frac{\ln 10}{z} \cdot (T-T_{ref})\right) \cdot \left(\frac{1}{1+C_c(t)}\right) \cdot dt$$

241

242 with $C_c(t)$ described by Eq (4)

$$\frac{dC_c}{dt} = -k_{\max} \cdot C_c$$

Similar to the pasteurisation/sterilisation/cooling principles (*P/F/C*-values) the mild heat value (*MH*-value) can be calculated in the following form when considering nonisothermal conditions and microbial parameter identification originates from microbial data that incorporate shoulder and/or tailing effects.

$$MH = \int_0^t \ln 10 \cdot \exp\left(\frac{\ln 10}{z} \cdot (T(t) - T_{ref})\right) \cdot \left(\frac{1}{1 + C_c(t)}\right) \cdot dt \quad (7)$$

248

$$\frac{dC_c}{dt} = -k_{\max}(T(t)) \cdot C_c(t) \quad (8)$$

249

250 Observe that the *MH*-value is given by a set of two equations in which the second 251 describes the evolution of the microbial physiological state of the cells. Under isothermal 252 conditions for an equivalent temperature T_{eq} Eqs (7), (8) will look as follows (the explicit 253 version is given).

- 254
- 255

$$MH = \ln 10 \cdot \exp\left(\frac{\ln 10}{z} \cdot (T_{eq} - T_{ref})\right) \cdot \left(1 + \frac{1}{\frac{\ln 10}{AsymD_{ref}}} \cdot \exp\left(\frac{\ln 10}{z} \cdot (T_{eq} - T_{ref})\right) \cdot 1\right) + \frac{1}{C_c(0) + 1} \cdot \left(1 + \frac{1}{\frac{\ln 10}{AsymD_{ref}}} \cdot \exp\left(\frac{\ln 10}{z} \cdot (T_{eq} - T_{ref})\right) \cdot 1\right) + \frac{1}{C_c(0) + 1} \right)$$
(9)

256

257

where instead of T(t) the equivalent temperature T_{eq} is chosen, in which the same microbial inactivation is achieved for the same time treatment (see Welt, Teixeira, Balaban, Smerage and Sage (1997) for a similar example when considering first order inactivation kinetics).

263 Microbial parameter identification

264

265 A so called global identification making use of all static experiments (presented in Figure 266 1) in one step was implemented. Parameters estimated taking into account the non-267 linearities are presented in Table 1. In the case that shoulder and tailing effects were not 268 considered, Eq. (3) reduces to the classical first order inactivation model. Similarly the 269 inactivation parameters were estimated for that case study and the results are illustrated in 270 Table 2. As expected, when assuming first order inactivation kinetics only for the log-271 linear portion of the inactivation kinetics the estimated microbial parameters i.e., 272 Asym D_{62} , z, coincide with those estimated during the non-log-linear regression analysis 273 (Table 3). The obtained z-values appear to be rather high but they seem to be on the range 274 of the estimated z-value ($z = 7^{\circ}$ C) of L. monocytogenes obtained for various food 275 products (van Asselt & Zwietering, 2006).

The conducted thermisation experiments resulted on the temperature profiles given in Figure 2. The targeting final temperature of 54 and 57°C was achieved in less than 20 minutes. The effect of non-log-linearity on describing the microbial inactivation kinetics was evaluated for both tested temperature profiles (by coupling these profiles with the microbial modelling models) (see Figure 2). These predictions are performed considering that inactivation of *L. innocua* is initiated at temperatures higher than 47°C (Valdramidis et al., 2008).

Depending on the severity of the temperature treatment the curves are diverging (see Figure 2). Considering that non-log-linearity described better the microbial data at hand it is evident that at temperature profiles targeting at 57°C an overestimation of the achieved microbial reductions is predicted from the classical log-linear modelling approaches (Figure 2). However for a target temperature of 54°C only when considering the microbial parameters estimated from a log-linear-regression of all the data results in the less conservative predictions for the achieved microbial reduction (Figure 2).

Further on, the developed methodological approach of the *MH*-value is tested for evaluating the efficiency of the dynamic temperature treatments when targeting at a specific microbial log reduction. Therefore, for a set of chosen equivalent temperatures, 293 i.e., 52, 55, 57°C, the *MH* -values are estimated by the use of Eq. (9) when considering 294 the three approaches of linearity and non-linearity. The desired microbial log reduction of 295 L. innocua was set to 4 logs (FIL-IDF, 1994) and 6 logs (safety margin (Claeys et al., 296 2002)). Observe that the derivation of *MH*-value (or the so called equivalent holding 297 time, t_{eq}) when considering log-linear inactivation were calculated by omitting the factor 298 which describes the physiological state of the cells in Eq. (9). Given that non-log-linearity 299 describes more accurately than log-linearity the static microbial inactivation data and 300 considering that the t_{eq} from the non-log-linear inactivation kinetics is the true one some 301 observations can be drawn from the results presented in Figure 3. On one hand when 302 assuming log-linear kinetics at the examined temperature range, i.e. case (iii), an over-303 processing treatment seems to happen which is more evident when targeting a 4 log 304 reduction. On the other hand when considering log-linear inactivation kinetics only for 305 the log-linear portion of the data, an under-processing effect is quite pronounced. These 306 results highlight that process efficiency can be wrongly calculated if assuming log-307 linearity for non-log-linear inactivation data. It should be noted that similar over-308 processing effect in case of log-linear inactivation kinetics at sterilisation temperatures 309 for microbial spores were observed by Corradini et al., (2006). In this study it appears 310 that the different types of log-linearity as well as the temperature range tested are both 311 influencing the estimated *MH* values. Inactivation temperature levels higher than 57° C 312 are expected to result in an interchange of the over- and under- processing regions (as the 313 lines depicted in Figure 3 are not parallel) while at much higher temperatures the 314 shoulder and tailing effects are less evident.



$$MH = AsymD_{ref} \cdot \int_0^t k(N, ,)dt \quad (11)$$

316

The advantages of this equation, if compared with previous literature studies, are twofold: (i) it takes into account non-log-linearity and (ii) it can easily be extended with respect to other environmental conditions or/and adjusted according to the microbial physiological state. Although similar concepts have been discussed in the literature like the method of Paired Equivalent Isothermal Exposures (PEIE) (Welt et al., 1997), they considered 322 irreversible first-order inactivation. Thus, in the case of microbial kinetics exhibiting a 323 shoulder effect, PEIE method would assume a straight line between two points on the 324 actual survivor curve. Nevertheless, there are recent studies that evaluate the non-log-325 linearity effects by assuming convex, concave or log-linear kinetics (Mafart et al., 2002; 326 Corradini et al., 2006; Sant'Ana et al., 2009). If these approaches are compared with the 327 developed *MH*-value then it can be seen that the parameters of the classical Bigelow 328 approach are retained in the current approach while the observed non-log-linearity, i.e., 329 shoulder, tailing effects are described by additional factors. Particularly, the non-log-330 linearity effect in those studies is described by the parameters δ and p and b and n, 331 respectively, which do not discriminate between log-linear and non-log-linear parts of the 332 inactivation kinetics. The advantage of the current approach is that the additional 333 parameters can be interpreted independently and Eq. (11) can be further specified 334 depending on the case-study at hand, for example, when more (environmental or 335 physiological) factors are considered.

336

337 **Future work**

338

339 Further studies on coupling microbial inactivation kinetics with heat transfer phenomena 340 especially for non-homogeneous products (including solid foods) are of interest for 341 designing similar thermal process. Investigation with other microorganisms and cases that 342 result in high non-log-linearity and comparison with the classical acceptable approaches 343 will work as additional validation of the developed modelling approach. For industrial 344 application purposes a comparative economical impact of the calculation of different 345 thermal death values is also of interest as it can avoid over and under-processing 346 schemes.

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443	Table 1. Parameter estimates (and their standard errors) derived from the non-log-linear
444	regression of Eqs 3-5 on the microbial data performed in capillary tubes at temperatures
445	58-66°C.

AsymD ₆₂ [min]	z [°C]	$\log(C_c(0)) [-]$	
0.32 ± 0.01	9.02 ± 0.24	1.14 ± 0.23	
log N(0) ₅₈ [-]	log N(0) ₆₀ [-]	$\log N(0)_{63}$ [-]	log N(0) ₆₆ [-]
8.99 ± 0.14	8.77 ± 0.14	9.01 ± 0.14	9.05 ± 0.16
logN _{res58} [-]	logN _{res60} [-]	$\log N_{res63}$ [-]	logN _{res66} [-]
2.89 ± 0.25	3.59 ± 0.14	2.77 ± 0.29	2.77 ± 0.25
	n antimatas (1-11-1	n standard america da '	d fuerre e l'arresta
Table 2. Paramete of the microbial da	r estimates (and their ata performed in capi	r standard errors) derive llary tubes at temperatur	ed from a linear regrees 58-66°C.
Table 2. Paramete of the microbial data $AsymD_{62}$ [min]	r estimates (and thei ata performed in capi z [°C]	r standard errors) derive llary tubes at temperatur	ed from a linear regreerer te solo from a linear regreerer te solo for the solo for
Table 2. Paramete of the microbial da Asym D_{62} [min] 0.41 ±0.13	r estimates (and their ata performed in capi z [°C] 8.75 ± 0.29	r standard errors) derive llary tubes at temperatur	ed from a linear regreerer te solo for the second sec
Table 2. Paramete of the microbial da Asym D_{62} [min] 0.41 ± 0.13 log $N(0)_{58}$ [-]	r estimates (and their ata performed in capi z [°C] 8.75 ± 0.29 $\log N(0)_{60}$ [-]	r standard errors) derive llary tubes at temperatur log <i>N</i> (0) ₆₃ [-]	ed from a linear regre res 58-66°C. log <i>N</i> (0) ₆₆ [-]
Table 2. Paramete of the microbial da Asym D_{62} [min] 0.41 ± 0.13 $\log N(0)_{58}$ [-] 9.19 ± 0.19	r estimates (and their ata performed in capit $z [^{\circ}C]$ 8.75 ± 0.29 $\log N(0)_{60} [-]$ 9.31 ± 0.24	r standard errors) derive llary tubes at temperatur $\log N(0)_{63}$ [-] 9.23 ± 0.19	ed from a linear regra res 58-66°C. $\log N(0)_{66}$ [-] 9.38 ± 0.26
Table 2. Paramete of the microbial da Asym D_{62} [min] 0.41 ± 0.13 log $N(0)_{58}$ [-] 9.19 ± 0.19	r estimates (and their ata performed in capi z [°C] 8.75 ± 0.29 $\log N(0)_{60}$ [-] 9.31 ± 0.24	r standard errors) derive llary tubes at temperatur $\log N(0)_{63}$ [-] 9.23 ± 0.19	ed from a linear regrates 58-66°C. $\log N(0)_{66}$ [-] 9.38 ± 0.26
Table 2. Paramete of the microbial da Asym D_{62} [min] 0.41 ± 0.13 log $N(0)_{58}$ [-] 9.19 ± 0.19	r estimates (and their ata performed in capi z [°C] 8.75 ± 0.29 $\log N(0)_{60}$ [-] 9.31 ± 0.24	r standard errors) derive llary tubes at temperatur $\log N(0)_{63}$ [-] 9.23 ± 0.19	ed from a linear regrates 58-66°C. $\log N(0)_{66}$ [-] 9.38 ± 0.26
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Figure 1. Regression analysis by taking into account shoulder and tailing effects (Left
Figure, continuous line), by assuming first order inactivation kinetics only for the loglinear portion of inactivation (Left Figure, dashed line), by assuming first order
inactivation kinetics for all inactivation data (Right Figure).



Figure 2. Dynamic temperature profile of a mild heat process representing thermisation with a target temperature of 54°C (Left), 57°C (Right) and associated microbial simulations (i) (considering non-log-linearity), continuous line, case (ii) (considering loglinearity when taking the log-linear portion of the data), dotted line, case (iii) (considering log-linearity for all data), dashed line,



Figure 3. Derivation of equivalent time and temperature values (o: considering loglinearity for all data), (Δ : considering log-linearity when taking the log-linear portion of the data), (x: considering non-log-linearity) when targeting at 4 log (top Figure) and 6 log reduction (bottom Figure) of *L. innocua*. Lines represent linear regression of the obtained data points (i) (considering non-log-linearity), continuous line, case (ii) (considering loglinearity when taking the log-linear portion of the data), dotted line, case (iii) (considering log-linearity for all data), dashed line, respectively.

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