

Predictive Microbiology

A tool to support food safety decisions

Models of microbial inactivation

- application in foods

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Escola Superior de Biotecnologia

7th December 2007

Mathematical modelling

some concepts ...

$$y = f(x, q) + e$$

Parameters estimation

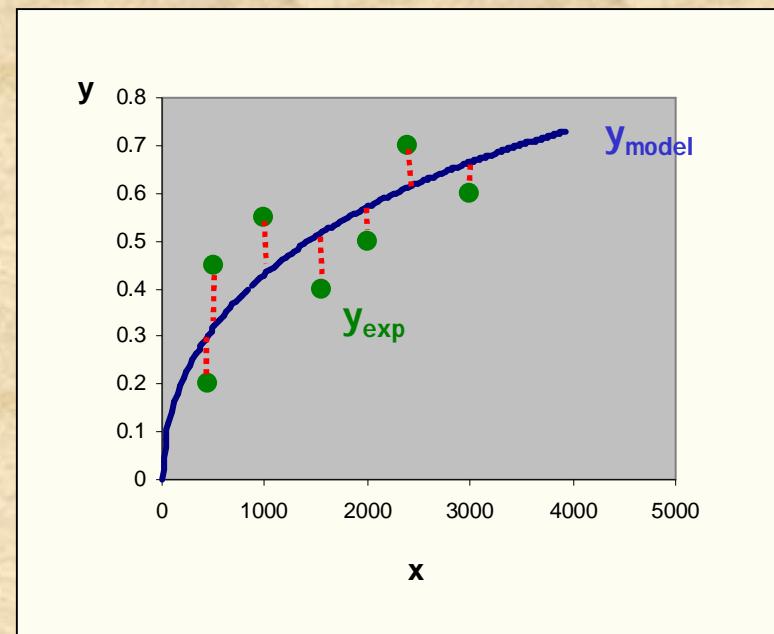
minimization of the residuals
between **experimental** values
and the ones predicted by the **model**



Precise ?

q^*

Accurate ?



Mathematical modelling

objective

precise and accurate description of data

model adequacy

quality of the parameters



Mathematical modelling

mechanistic models

- ◆ fundamental description of the processes involved
- ◆ more complex

empirical models

- ◆ *black box*
- ◆ more simple (or not!)
- ◆ practical application



advantages and **disadvantages** should be considered
decision depending on the **final goal**

Mathematical modelling

mathematical complexity



model adequacy

model

quality

parameters



advantages

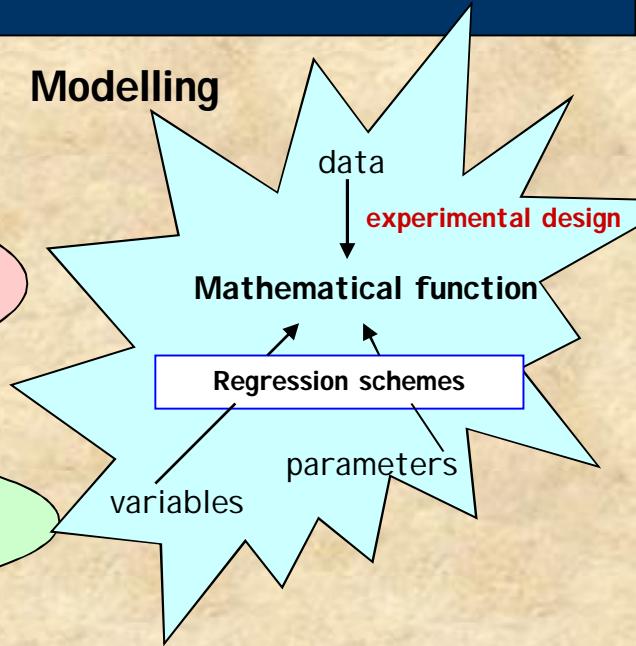
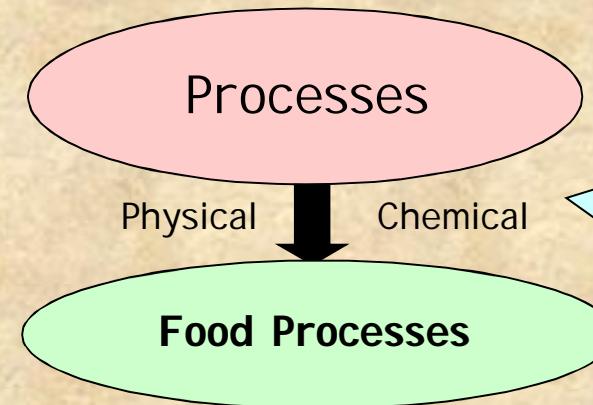
- knowledge of the process
- process effect in the product
- control of the variables involved

Mathematical modelling

Transport Phenomena
• heat
• mass
• *momentum*

Reaction kinetics

Properties

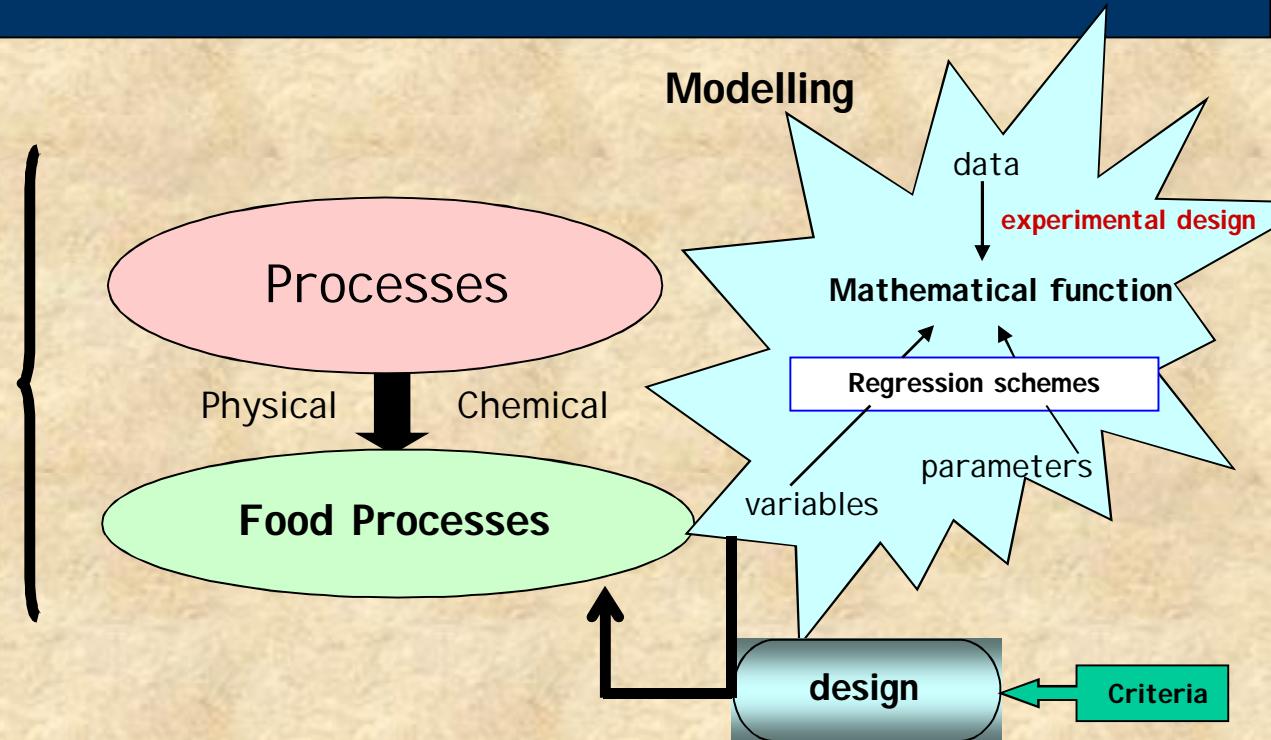


Mathematical modelling

Transport Phenomena
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Reaction kinetics

Properties

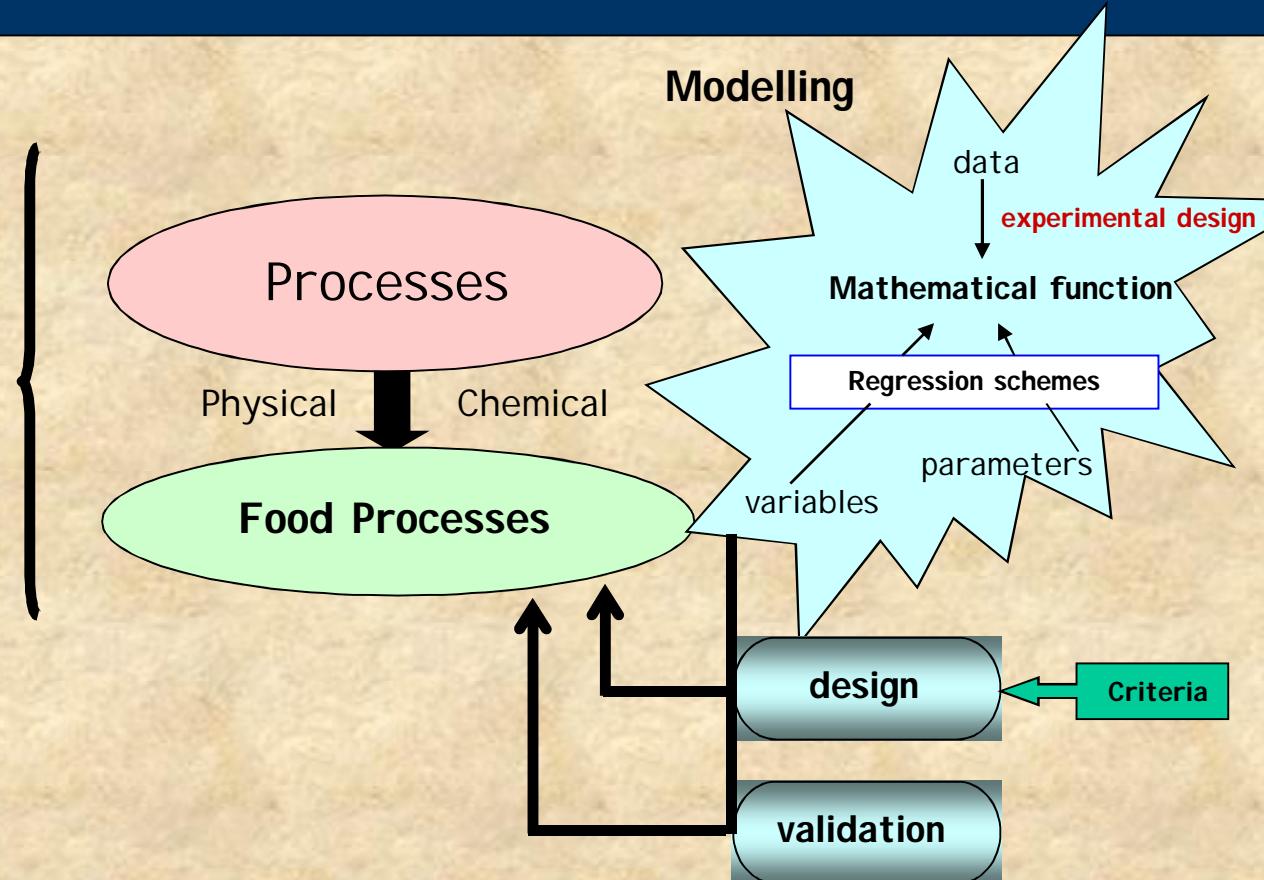


Mathematical modelling

Transport Phenomena
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Properties

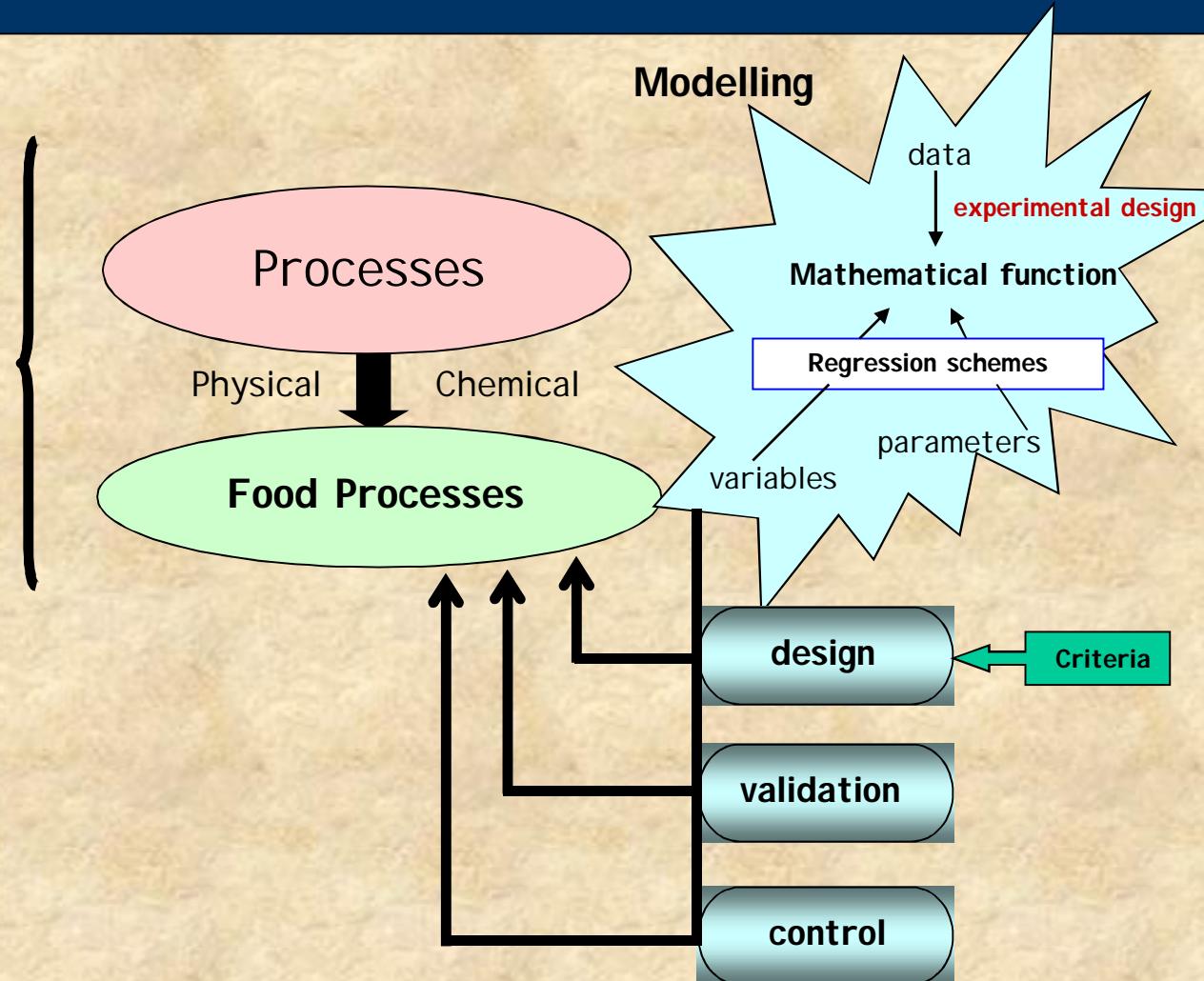


Mathematical modelling

Transport Phenomena
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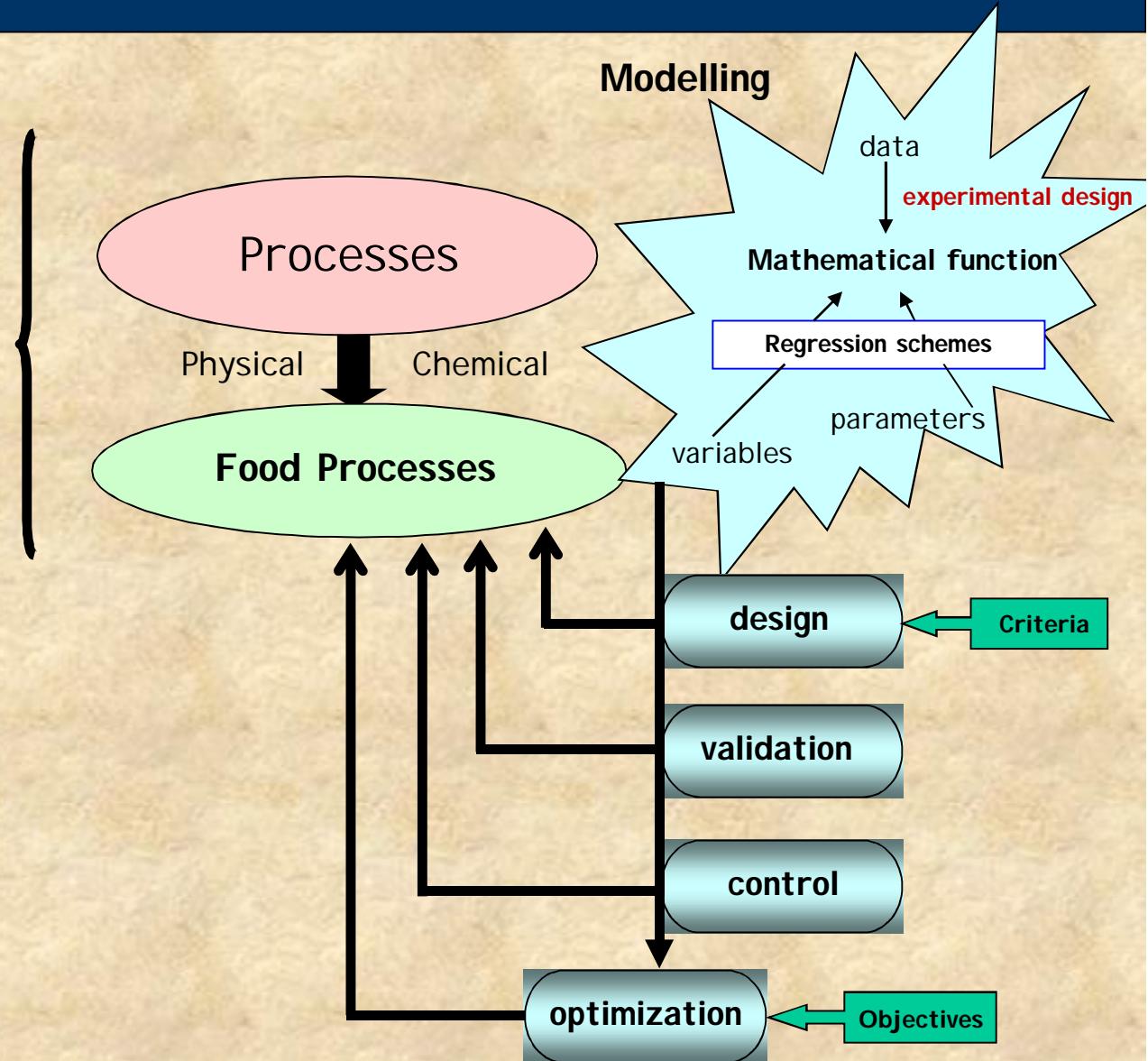


Mathematical modelling

Transport Phenomena
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Reaction kinetics

Properties

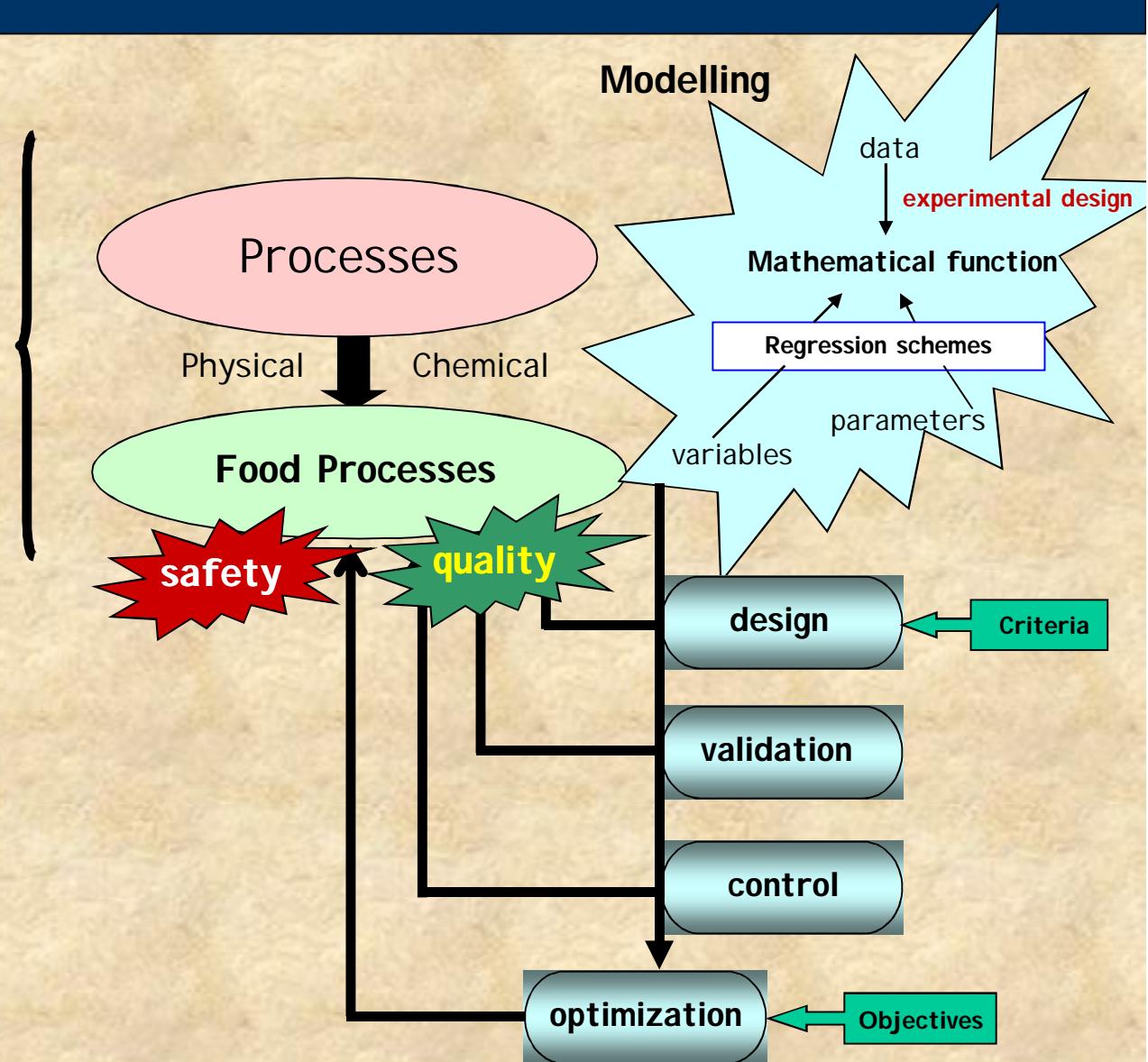


Mathematical modelling

Transport Phenomena
• heat
• mass
• *momentum*

Reaction kinetics

Properties



Mathematical modelling



safety

Mathematical modelling

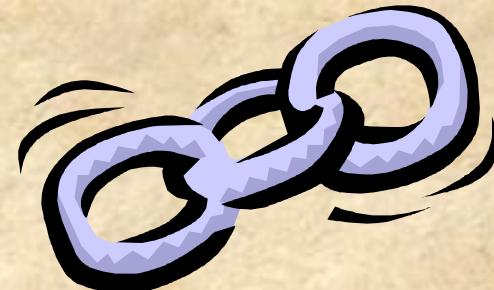


**predictive
microbiology**

Mathematical modelling



microbiology



mathematics

statistics

application

- prediction / simulation
- development of efficient
inactivation processes

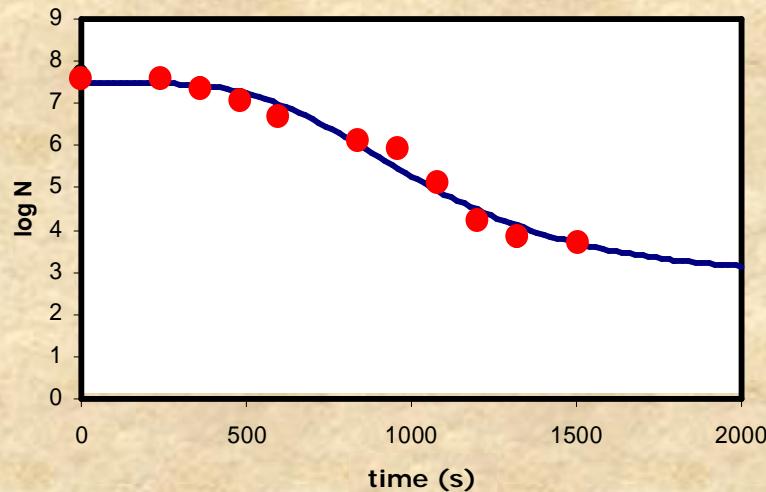


contribution to safety

inactivation

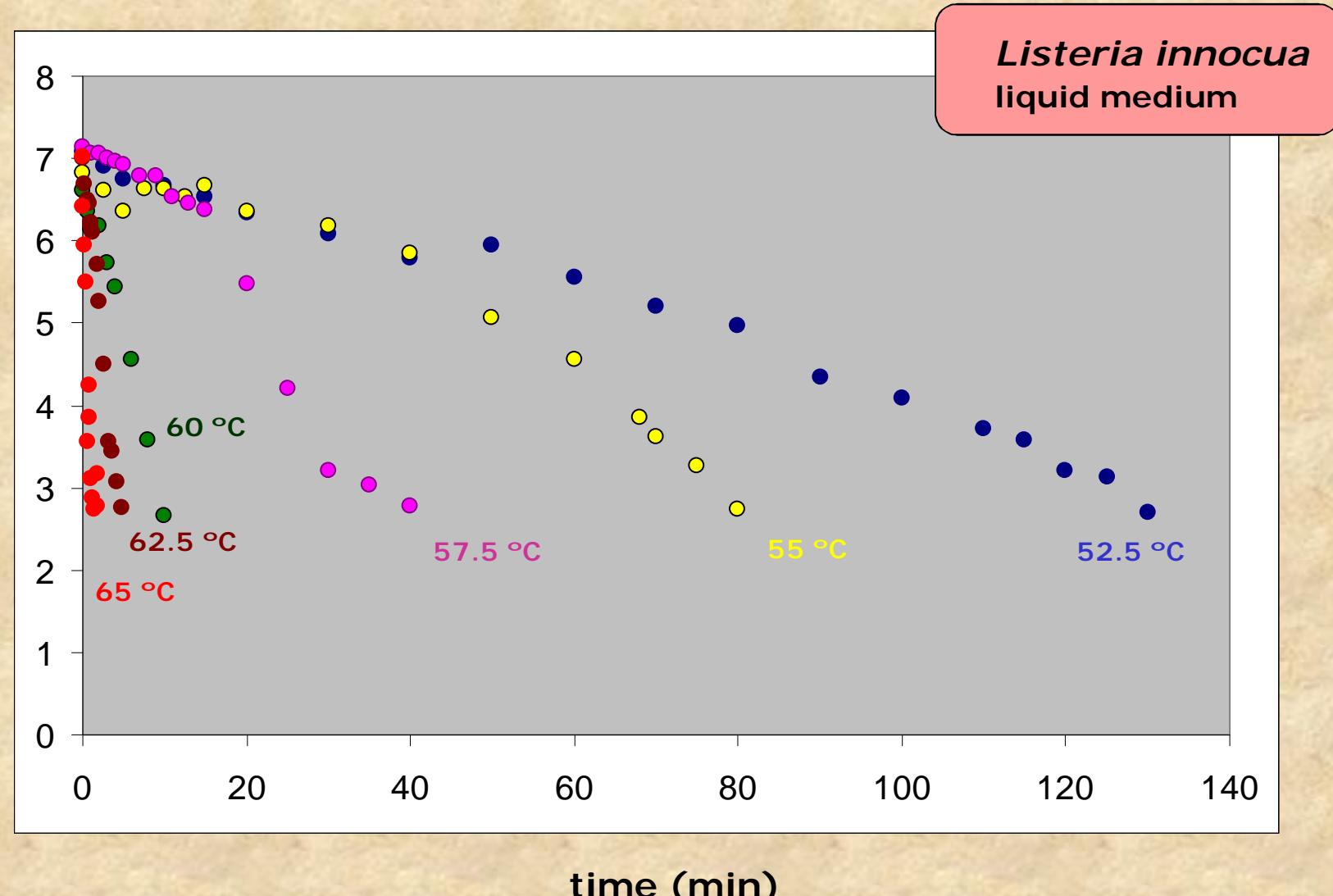


Sigmoidal behaviour



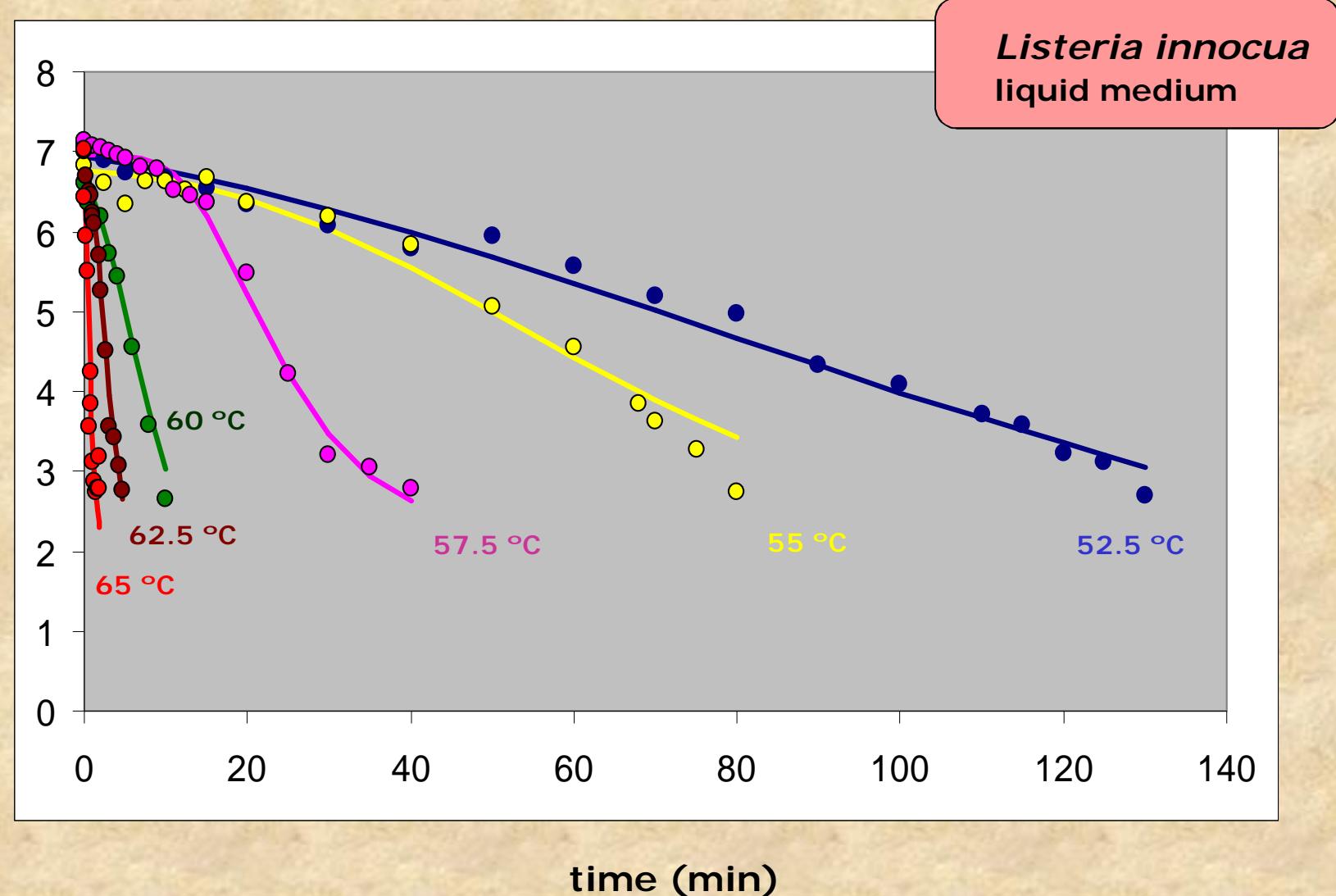
Reflects the presence of microbial sub-populations
more **temperature resistant**
(or other environmental **stressing factor**)

Example



Miller (2004)

Example



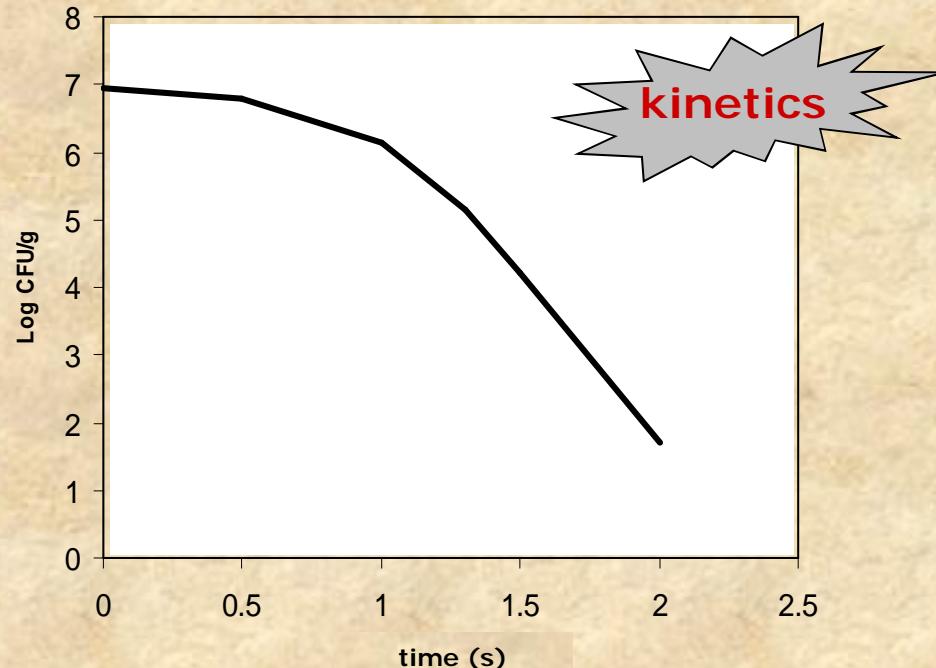
Miller (2004)

Mathematical models

❖ primary



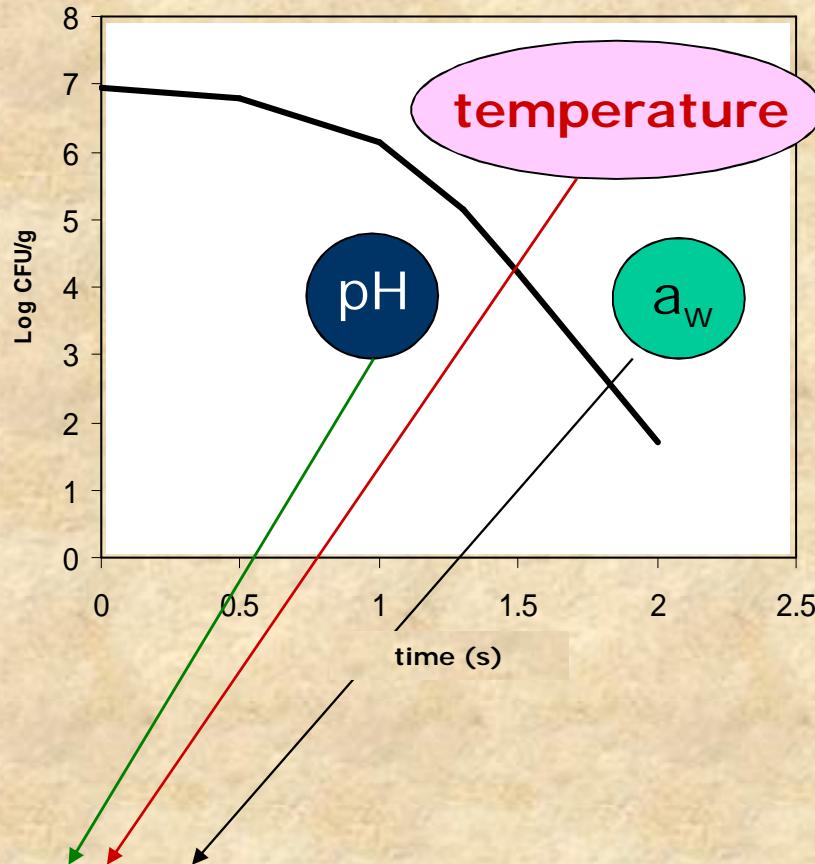
parameters



Mathematical models

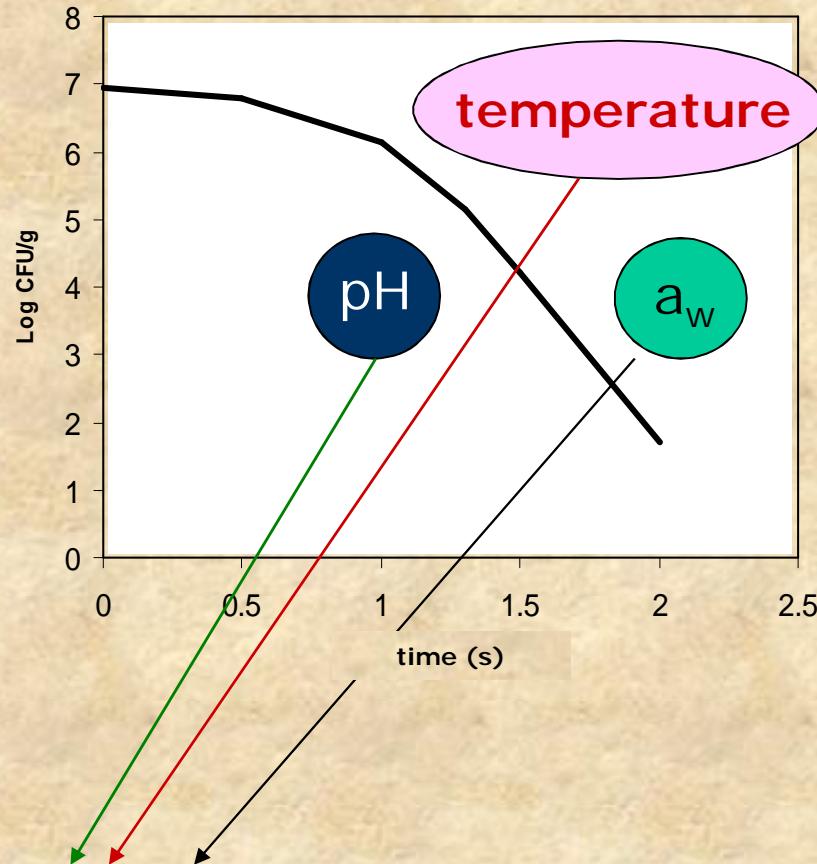
❖ primary

❖ secondary
parameters



Mathematical models

❖ primary

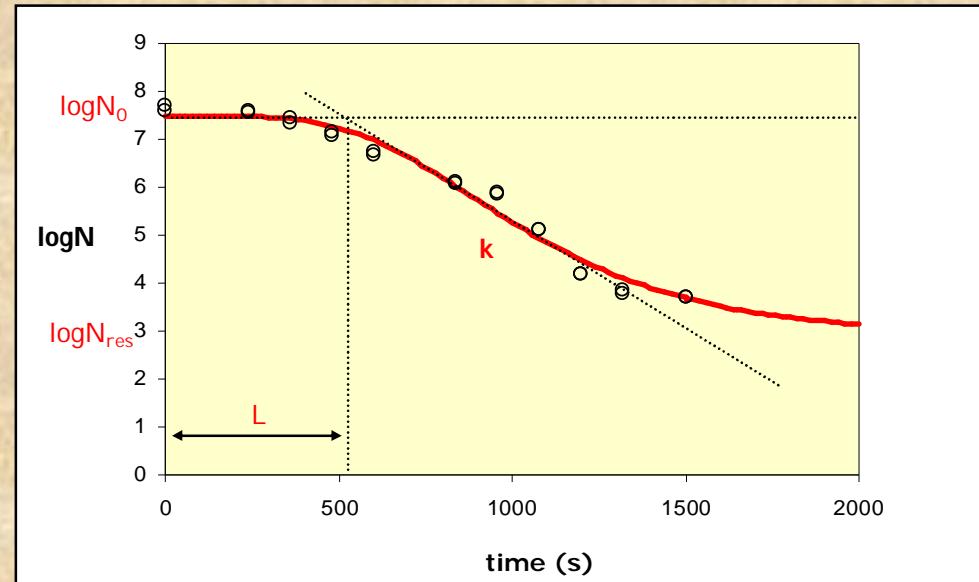


❖ secondary
parameters

❖ tertiary - all models integration - software

Models for inactivation

❖ primary

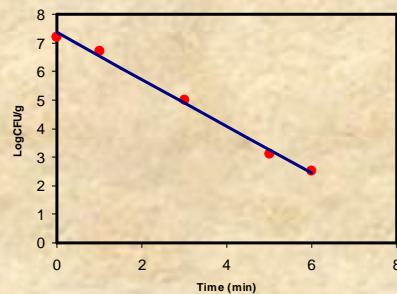


empirical

fundamental

Models for inactivation

❖ primary



$$N = N_0 \exp(-kt)$$



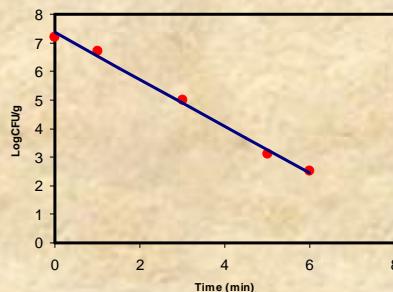
$$\log N = \log N_0 - \frac{t}{D}$$

1st order

D – decimal reduction time

Models for inactivation

❖ primary



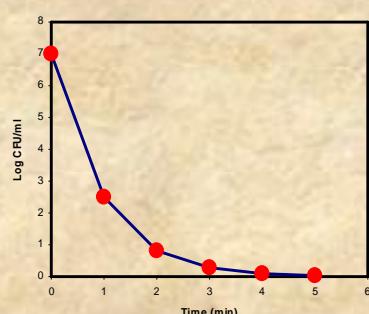
$$N = N_0 \exp(-kt)$$



$$\log N = \log N_0 - \frac{t}{D}$$

1st order

D – decimal reduction time



biphasic

$$\frac{N}{N_0} = F_1 \exp(-k_1 t) + (1 - F_1) \exp(-k_2 t)$$

Cerf
(1977)

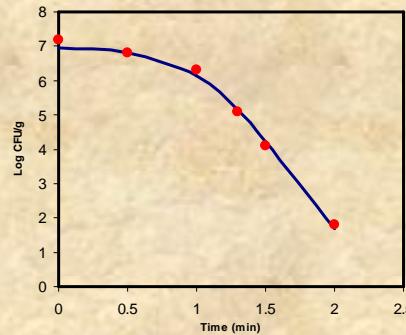
F_1 – fraction of inactivated microorganisms
 k_1 e k_2 – kinetic constants

$$\log \frac{N}{N_0} = \log \left(\frac{2F_1}{1 + \exp(k_1 t)} + \frac{2(1 - F_1)}{1 + \exp(k_2 t)} \right)$$

Kamau et al.
(1990)

Models for inactivation

❖ primary



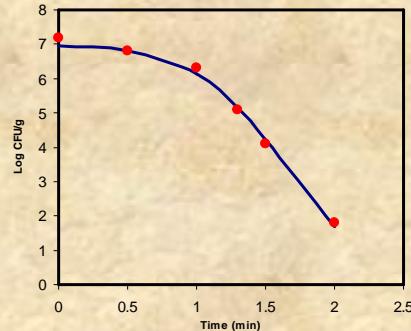
Whiting & Buchanan
(1992)

$$\log \frac{N}{N_0} = \log \left(\frac{F_1(1 + \exp(-k_1L))}{1 + \exp(k_1(t - L))} + \frac{(1 - F_1)(1 + \exp(-k_2L))}{1 + \exp(k_2(t - L))} \right)$$

L – lag or shoulder

Models for inactivation

❖ primary



Whiting & Buchanan
(1992)

$$\log \frac{N}{N_0} = \log \left(\frac{F_1(1 + \exp(-k_1L))}{1 + \exp(k_1(t - L))} + \frac{(1 - F_1)(1 + \exp(-k_2L))}{1 + \exp(k_2(t - L))} \right)$$

L – lag or shoulder

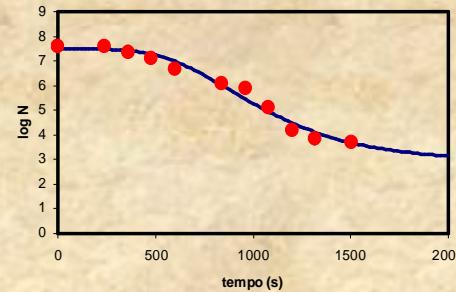
Cole et al.
(1993)

$$\log N = \alpha + \frac{w - \alpha}{1 + \exp\left(\frac{4\sigma(\lambda - \log t)}{w - \sigma}\right)}$$

Distribution of
microbial population
heat sensitiveness

Models for inactivation

❖ primary



Baranyi et al.
(1993)

$$\frac{dN}{dt} = -k \alpha(t) \beta(t) N$$

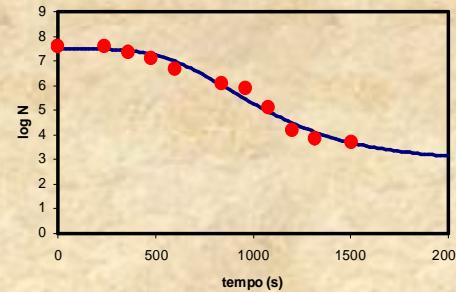
$$N(t = 0) = N_0$$

function 'tail'

function 'lag'

Models for inactivation

❖ primary



Baranyi et al.
(1993)

$$\frac{dN}{dt} = -k \alpha(t) \beta(t) N$$

$$N(t = 0) = N_0$$

function 'tail'

function 'lag'

Geeraerd et al.
(2000)

$$\frac{dN}{dt} = -k_{\max} k_Q(Q) N$$

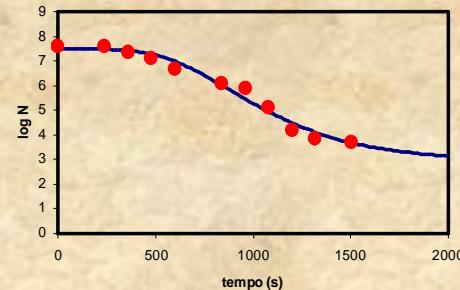
$$\frac{dQ}{dt} = -k_{\max} Q$$

$$\log\left(\frac{N}{N_0}\right) = \log(\exp(-k_{\max}t)) \frac{1 + Q(0)}{1 + Q(0)\exp(-k_{\max}t)}$$

Q – variable related to the physiological state of microorganisms

Models for inactivation

❖ primary



Gompertz

Bhaduri et al (1991)
Linton et al. (1995, 1996)
Xiong et al. (1999)

↓
Listeria monocytogenes

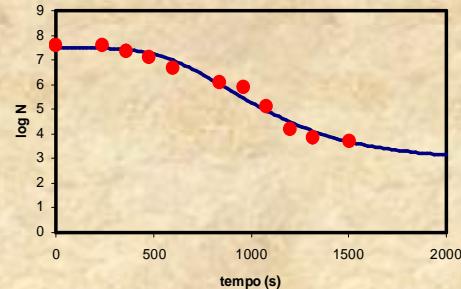
$$\log N = \log N_0 + \log\left(\frac{N_{res}}{N_0}\right) \exp\left(-\exp\left(-\frac{k e}{\log\left(\frac{N_{res}}{N_0}\right)}(L - t)\right) + 1\right)$$



Reparameterization for inactivation based on Zwitering et al. (1990)

Models for inactivation

❖ primary



Gompertz

Bhaduri et al (1991)
Linton et al. (1995, 1996)
Xiong et al. (1999)

Listeria monocytogenes

$$\log N = \log N_0 + \log\left(\frac{N_{res}}{N_0}\right) \exp\left(-\exp\left(-\frac{k e}{\log\left(\frac{N_{res}}{N_0}\right)}(L - t) + 1\right)\right)$$



Reparameterization for inactivation based on Zwitering et al. (1990)

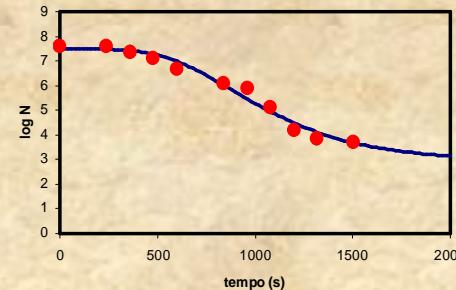
Logistic

$$\log N = \frac{c}{1 + \exp(k(t - L))}$$

c – constant

Models for inactivation

❖ Gompertz



elected in our research

$$\log N = \log N_0 + \log\left(\frac{N_{res}}{N_0}\right) \exp\left(-\exp\left(-\frac{k e}{\log\left(\frac{N_{res}}{N_0}\right)}(L - t) + 1\right)\right)$$

or

$$\log \frac{N}{N_0} = \log\left(\frac{N_{res}}{N_0}\right) \exp\left(-\exp\left(-\frac{k e}{\log\left(\frac{N_{res}}{N_0}\right)}(L - t) + 1\right)\right)$$

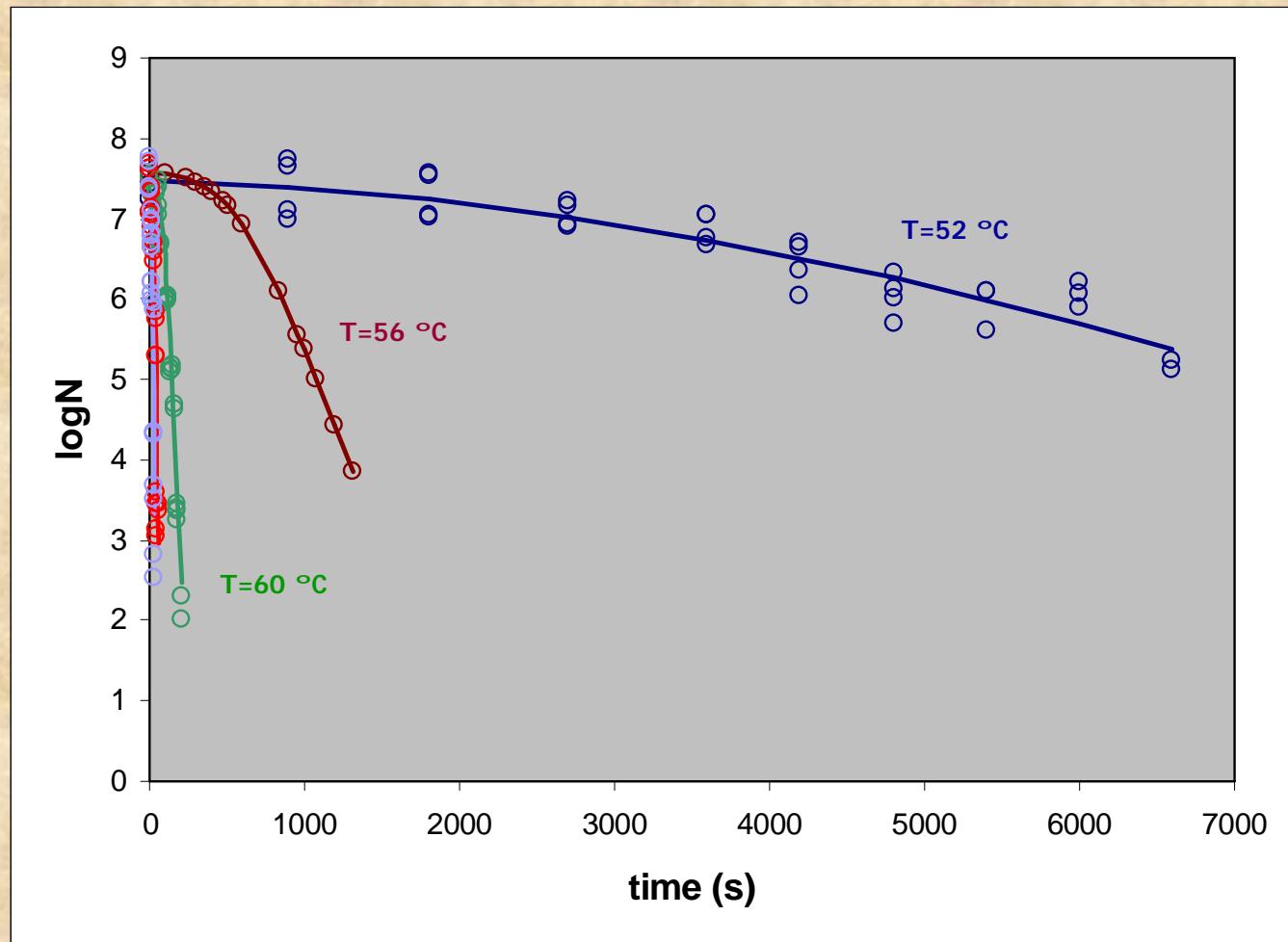
Examples

Data of *L. monocytogenes* Scott A at 52,56,60,64,68°C

(24 hours incubation at 5°C in half cream)

Casadei et al. (1998)

Gompertz



Gil (2002)

Statistica 6.0

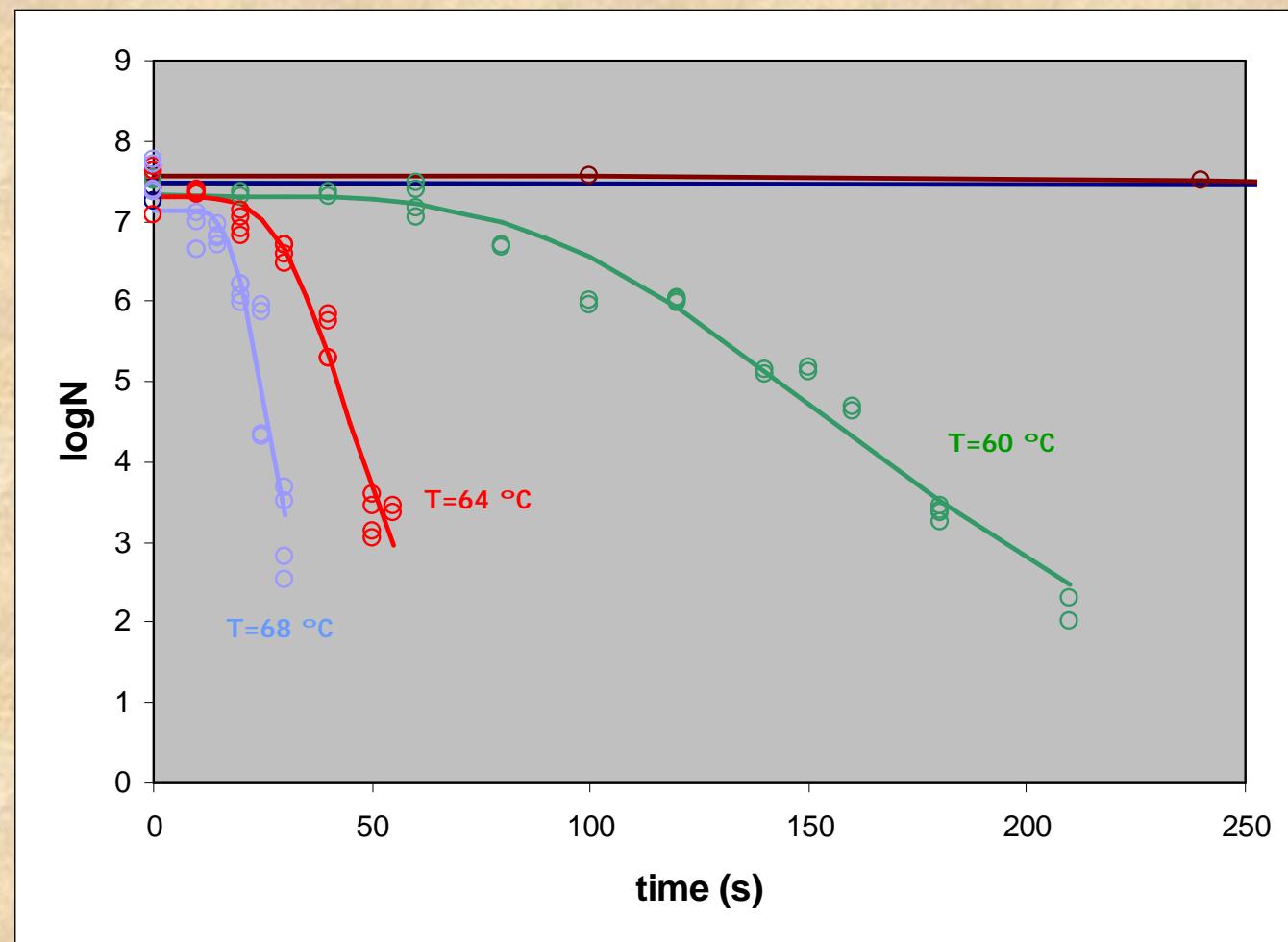
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Casadei et al. (1998)

Gompertz

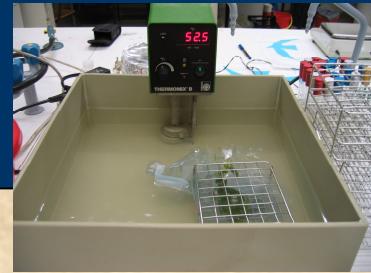


Gil (2002)

Statistica 6.0

Examples

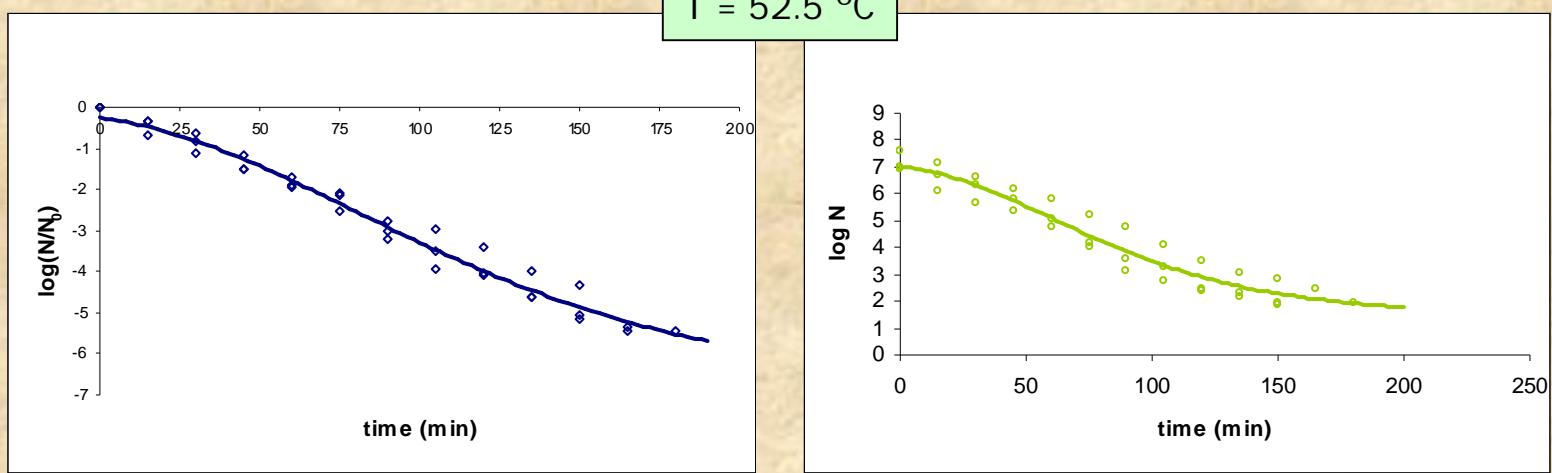
Data of *L. innocua* at 52.5 and 60°C
Parsley (*Petroselinum crispum*) artificially contaminated



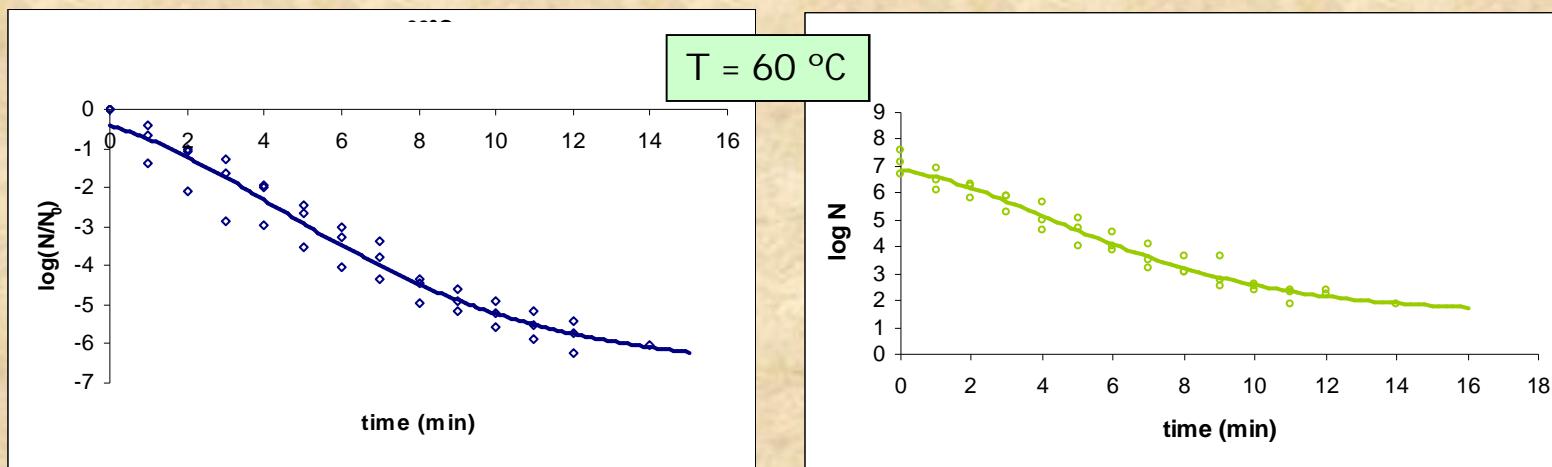
Gompertz

$$\log \frac{N}{N_0}$$

T = 52.5 °C



Miller (2006)

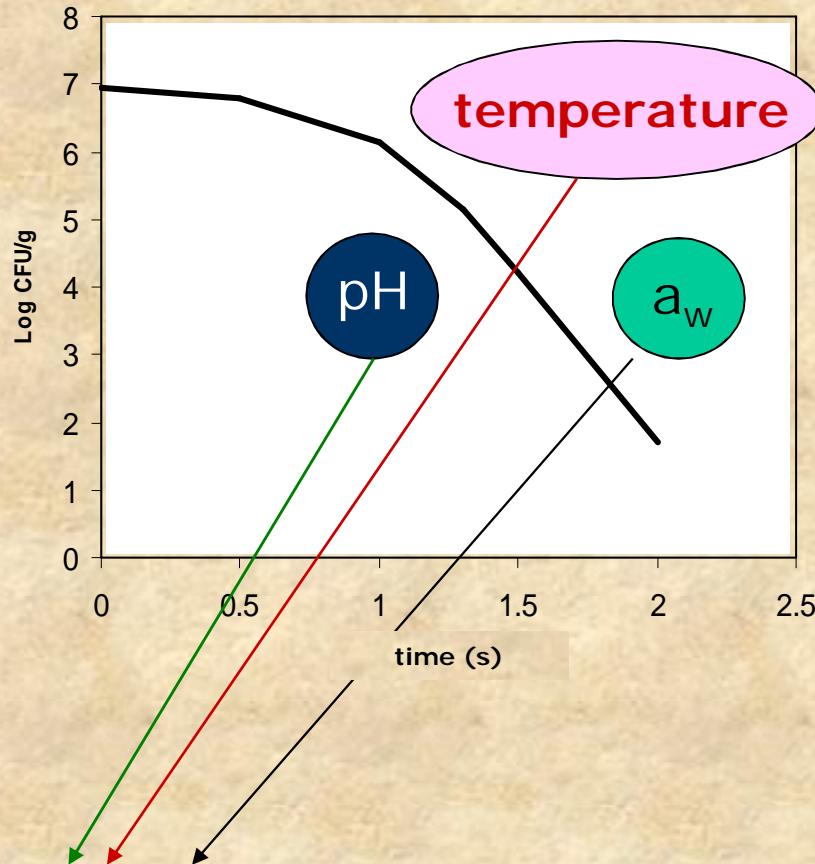


Statistica 6.0

Mathematical models

❖ primary

❖ secondary
parameters



Mathematical models

❖ secondary

Arrhenius

$$k = k_0 \exp\left(-\frac{E_a}{RT}\right) \rightarrow \ln k = \ln k_0 - \frac{E_a}{RT}$$

$$k = k_{ref} \exp\left(-\frac{E_a}{R}\left(\frac{1}{T} - \frac{1}{T_{ref}}\right)\right)$$

Davey / Arrhenius modified

$$\ln k = C_0 + \frac{C_1}{T} + \frac{C_2}{T^2} + C_3 a_W + C_4 a_W^2$$

“Square-root type models”

Ratkowsky *et al.* (1982)

$$\sqrt{k} = b(T - T_{min})$$

McMeekin *et al.* (1987)

$$\sqrt{k} = b(T - T_{min}) \sqrt{(a_w - a_{wmin})}$$

Adams *et al.* (1991)

$$\sqrt{k} = b(T - T_{min}) \sqrt{(pH - pH_{min})}$$

McMeekin et al. (1992)

$$\sqrt{k} = b(T - T_{min}) \sqrt{(a_w - a_{wmin})} \sqrt{(pH - pH_{min})}$$

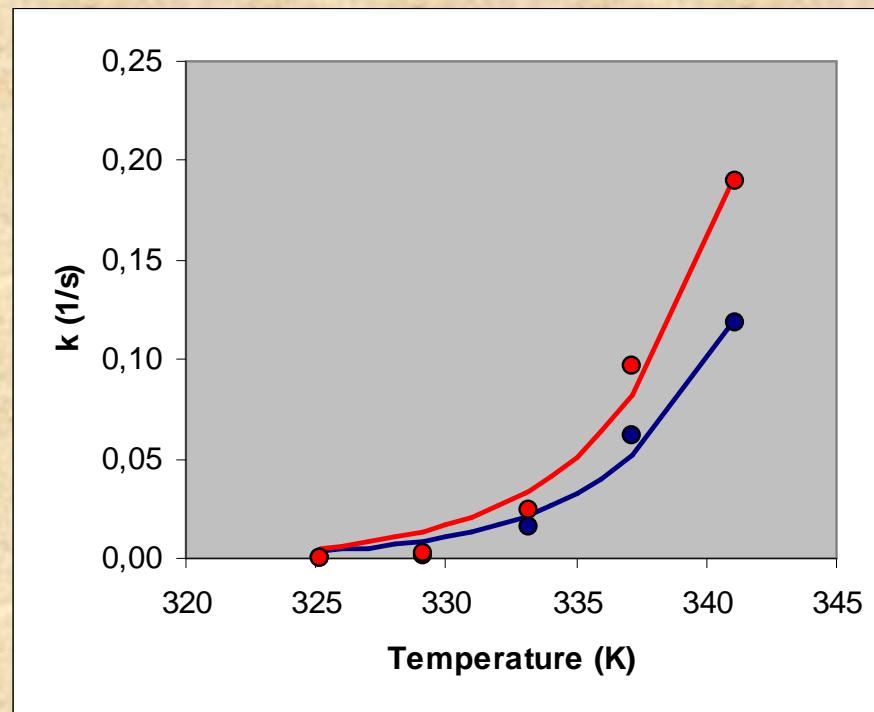
min – minimal value for growth

Examples

Data of *L.monocytogenes* Scott A
(24 hours incubation at 5°C in half cream)

Arrhenius

$k = f(T)$



Gompertz $k=0.0216 \exp(-203.3/R*(1/T-1/333.15))$

$$\left| \begin{array}{l} DE_a=28.85 \text{ kJ/mol} \\ DK_{ref}=4.58 \times 10^{-3} \text{ s}^{-1} \end{array} \right.$$

Logistic $k=0.0337 \exp(-206.6/R*(1/T-1/333.15))$

$$\left| \begin{array}{l} DE_a=27.56 \text{ kJ/mol} \\ DK_{ref}=7.31 \times 10^{-3} \text{ s}^{-1} \end{array} \right.$$

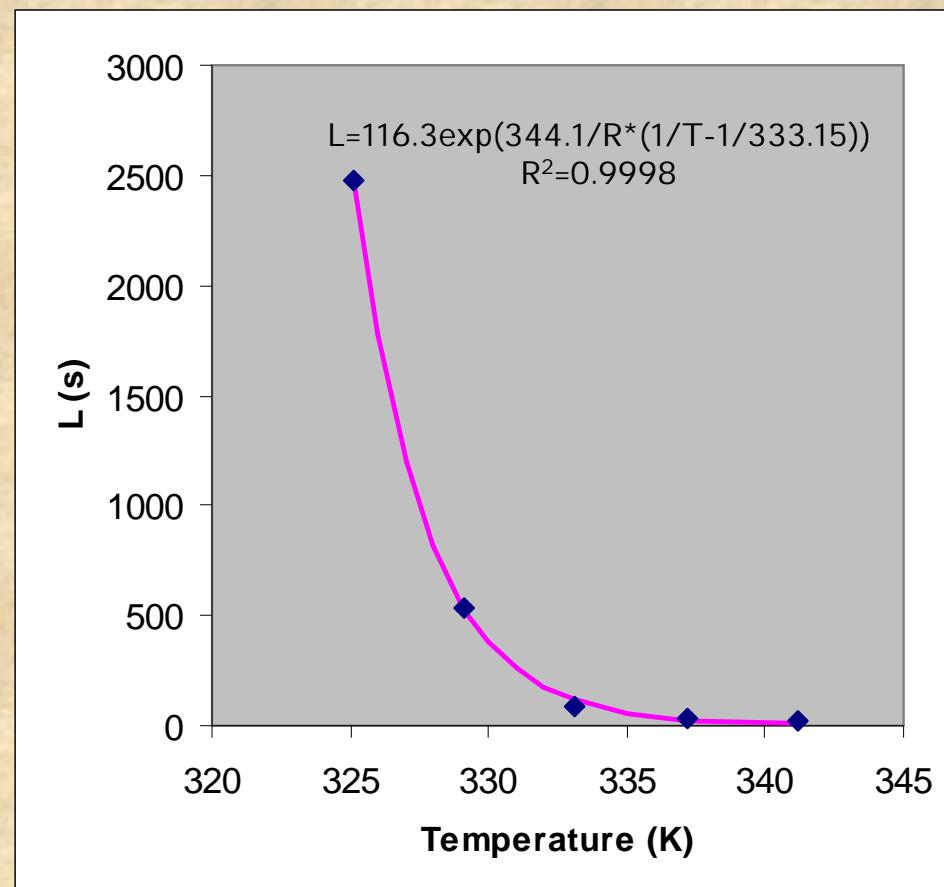
Examples

Data of *L.monocytogenes* Scott A
(24 hours incubation at 5°C in half cream)

Gompertz

$$DE_a = 7.485 \text{ kJ/mol}$$

$$DL_{ref} = 7.595 \text{ s}^{-1}$$



lag = f(T)

Examples

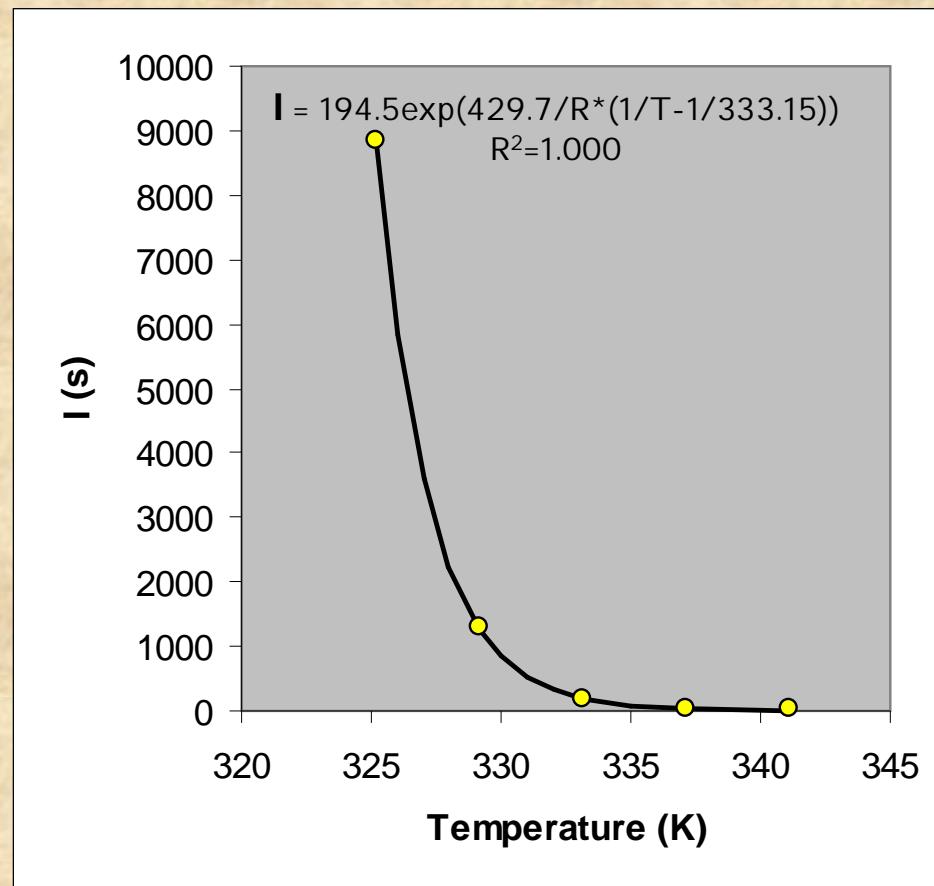
Data of *L.monocytogenes* Scott A
(24 hours incubation at 5°C in half cream)

Logistic

lag = f(T)

$$DE_a = 3.591 \text{ kJ/mol}$$

$$DI_{ref} = 6.154 \text{ s}^{-1}$$



Mathematical models

time-varying conditions



temperature

pH

a_w

More complexity !!



$$\frac{d(\log N)}{d(\text{tempo})}$$

Mathematical models

Gompertz

time-varying temperature conditions

$$\downarrow \frac{d(\log N)}{d(\text{tempo})}$$

$$\log N = \log N_0 + \int_0^t \left[k \exp(1) \exp \left(-\frac{k \exp(1)}{\log \left(\frac{N_{\text{res}}}{N_0} \right)} (L - t') + 1 \right) \exp \left(-\exp \left(-\frac{k \exp(1)}{\log \left(\frac{N_{\text{res}}}{N_0} \right)} (L - t') + 1 \right) \right) \right] dt'$$

$$k = k_{\text{ref}} \exp \left(-\frac{E_a}{R} \left(\frac{1}{T} - \frac{1}{T_{\text{ref}}} \right) \right)$$

$$L = a \exp \left(b \left(\frac{1}{T} - \frac{1}{T_{\text{ref}}} \right) \right)$$

Mathematical models

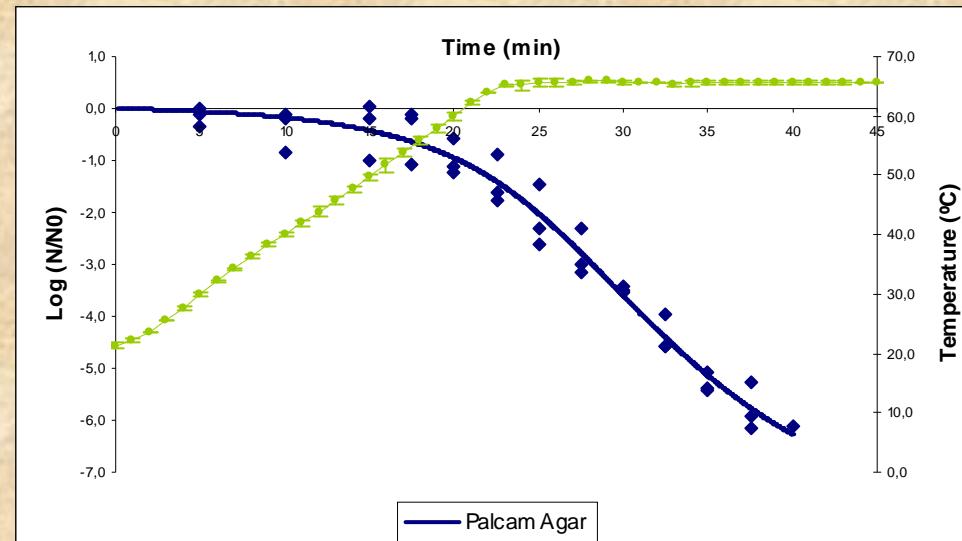
Parsley (*Petroselinum crispum*) artificially contaminated with *L. innocua*

Gompertz



Miller (2006)

time-varying temperature conditions



temperature increase rate ~ $2 \text{ } ^\circ\text{Cmin}^{-1}$

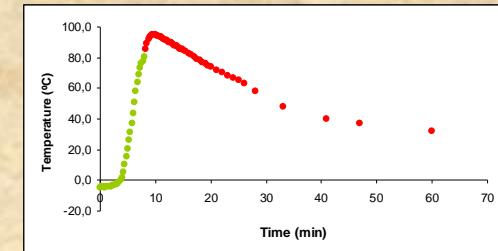
Mathematical models

Meat pockets artificially contaminated with *L. innocua*

Gompertz

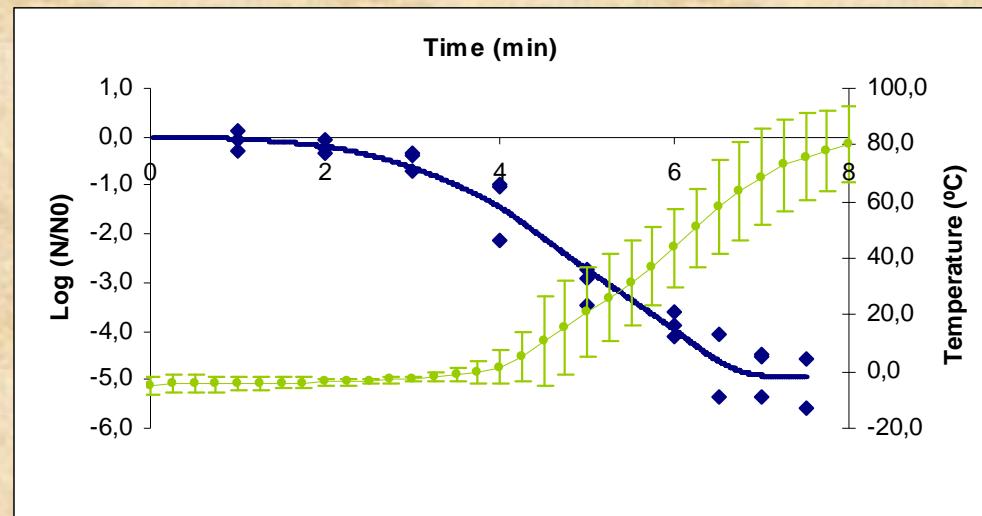


frying process



Miller (2006)

time-varying temperature conditions



maximum temperature increase rate $\sim 20 \text{ }^{\circ}\text{Cmin}^{-1}$

Models for inactivation

❖ tertiary

softwares

Microbial growth

Shelf life prediction

Microbial inactivation



lacunes

Drawbacks

- microbial interaction
- natural strains diversity
- complexity of food structure
- food/microorganism
- modelling 'lag'
- modelling 'tail'
- predictions in real time-varying environmental conditions

Participation of our research team in the project ...

2001-2004

BUGDEATH

www.frperc.bris.ac.uk/bugdeath.htm

Predictive
microbiology

Surface
pasteurization of
foods

important boundary in food contamination

"BUGDEATH" (QLRT-2001-01415)



Micrograph of a *Salmonella* bacteria

Predicting the reduction in microbes on the surface of foods during pasteurisation.

- [Tell me more about the Bugdeath project.](#)
- [What are the objectives of the Bugdeath project?](#)
- [Who are the project partners?](#)
- [Meetings](#)
- [Industrial Advisory Group \(IAG\)](#)
- [Partner area of web site \(specifically for project partners - password needed to enter\)](#)

Bugdeath

BUGDEATH is a research project funded by the European Commission to produce accurate predictive models of the reductions in microbial numbers that can be achieved on the surface of foods during surface pasteurisation processes.

Food poisoning is increasing throughout the European Union (EU). Over 60% of outbreaks are associated with meat, fresh fruit and salad vegetables. Most of the contamination by pathogenic and spoilage organisms is present on the surface of foods at the time of harvesting or is transferred to the surfaces during slaughter and processing. Accurate microbial death models would be of considerable help to the food industry in the development of surface pasteurisation systems for meat, fruit and vegetables. This will in turn lead to safer foods with improved quality and shelf life.



Personnel involved in Bugdeath from left to right:

Partners

1. Food Refrigeration and Process Engineering Research Center (**FRPERC**)
University of Bristol
2. Escola Superior de Biotecnologia (**ESB**)
Universidade Católica Portuguesa
3. Department of Chemical Engineering
Katholieke Universiteit Leuven (**KUL**)
4. Teagasc, The National Food Center (**NFC**)
5. Campden & Chorleywood Food Research Association (**CCFRA**)
6. Faculty of Applied Science
University of West England (**UWE**)
7. Laboratoire de Génie des Procédés Alimentaires (**ENITIA**)
Ecole Nationale d'Ingenieurs des Techniques des Industries Agricoles et Alimentaires
8. Institute National de la Recherche Agronomique (**INRA**)

Objectives

- microbial inactivation studies at food surfaces
- development of precise and accurate inactivation models
- development of an equipment for surface food pasteurization
'rig apparatus'
- software development for prediction

ESB task

Foods under study

potato

chicken

meat

microorganisms

Listeria monocytogenes

E. coli

Salmonella

Campylobacter

rig apparatus

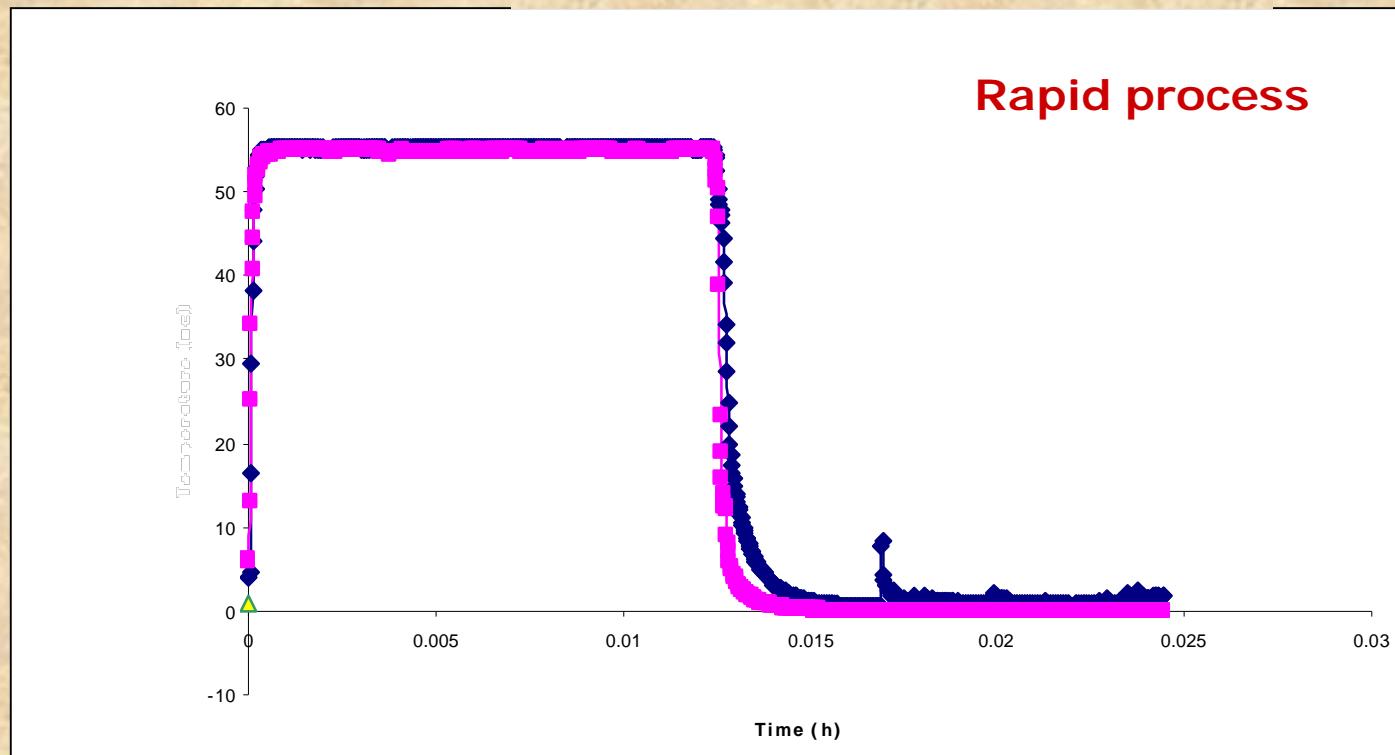


Bugdeath

Thermal processes

- steam
- dry air

Temperature histories



rig apparatus

Rapid process

Different
temperature
histories can
be tested

Team

Pedro Pereira
Maria Manuel Gil
Teresa Brandão
Cristina Silva

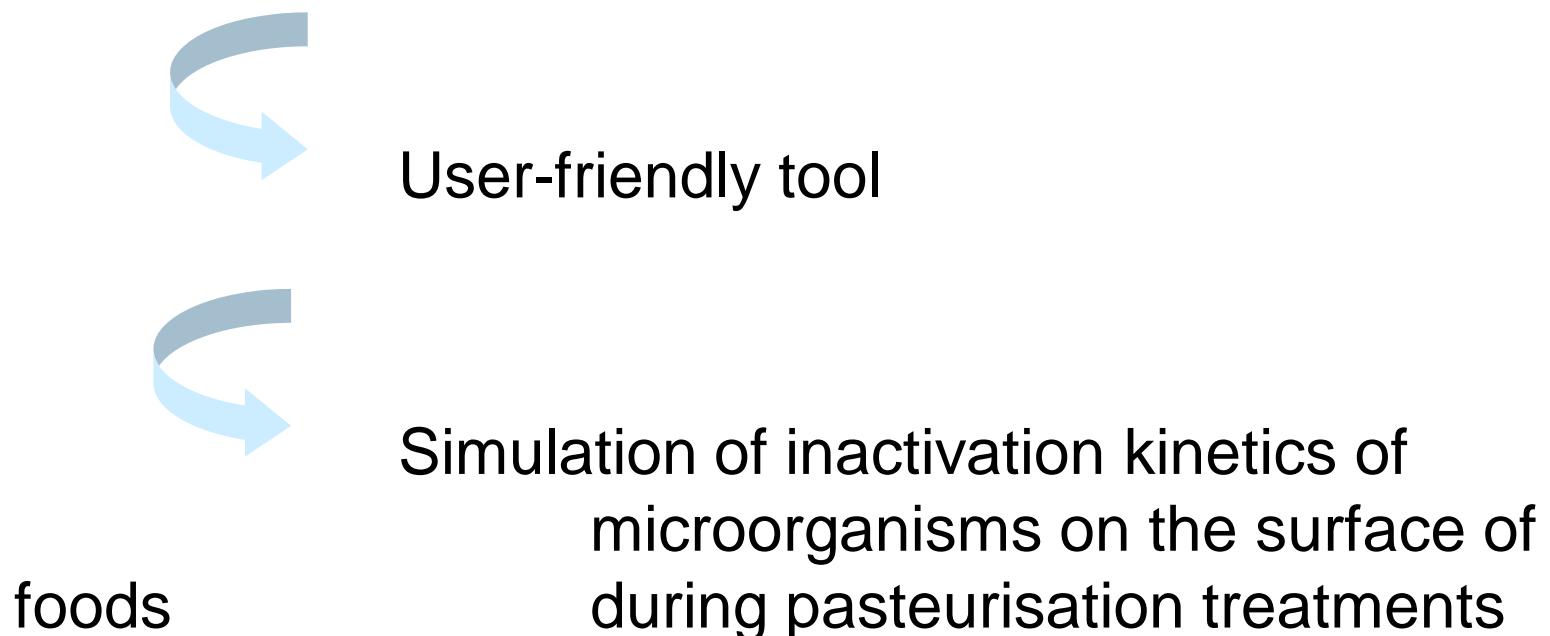


Dissemination session
Chipping Campden, Agosto 2004

Development of a user-friendly combined heat, mass transfer and microbial death model

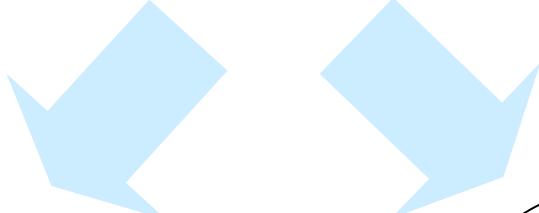
Objectives

- Create a software application



Introduction

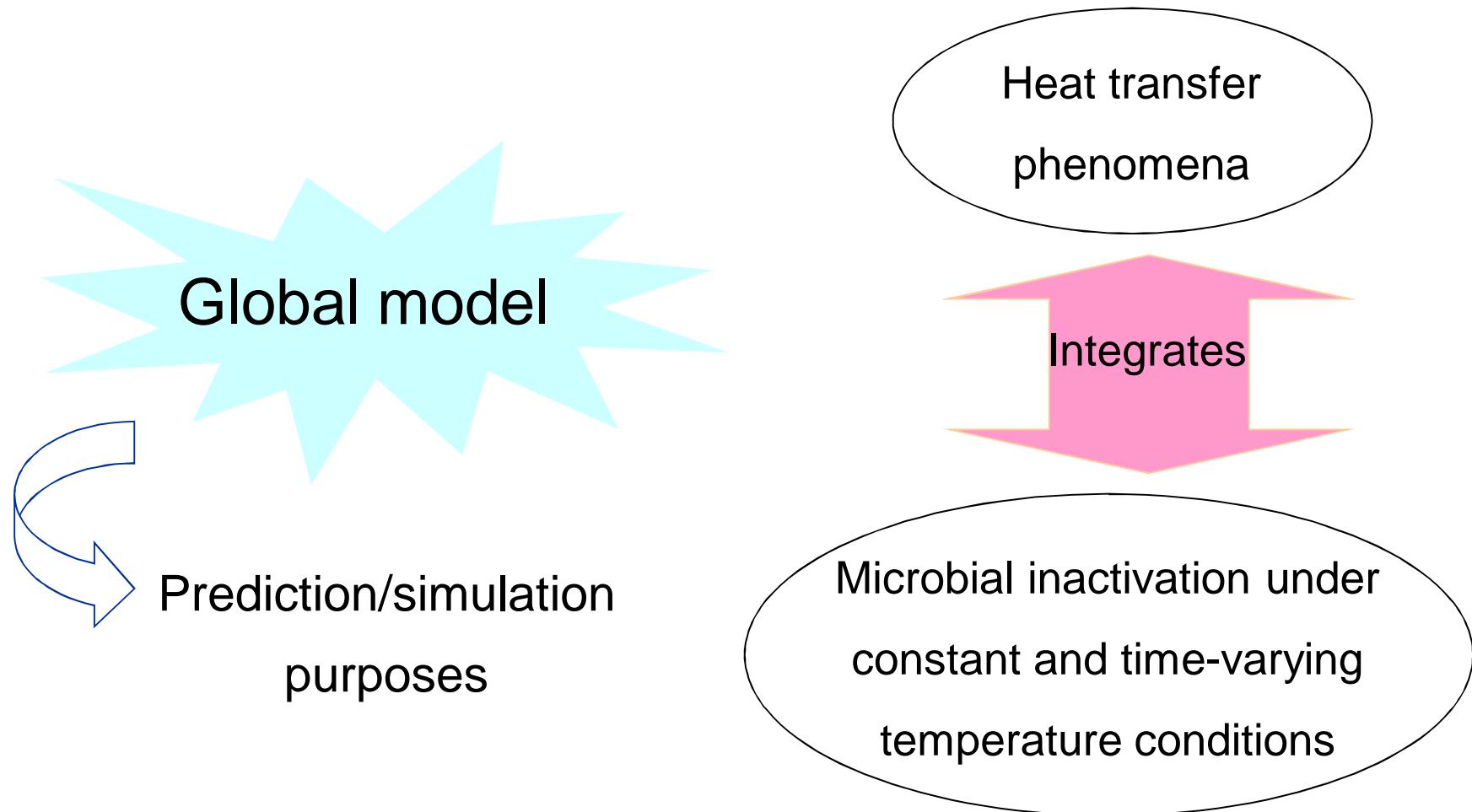
- Need to obtain reliable data on relationship between microbial death and the surface temperature of real foods



Lead to the design,
construction and
commission of equipment

Software application to
simulate results obtained in the
rig apparatus

Modelling



'Bugdeath' - funded by the European Commission under the EC Framework 5;
Quality of Life and Management of Living Resources Programme



Modelling

two modelling approaches

- Accurate modelling of heat transfer
 - Description of the phenomena induced to the food surface by a thermal process
- Modelling microbial inactivation behaviour under such temperature conditions

Heat transfer model

estimation of surface temperature

- One dimensional heat transfer model
- Combination of different phenomena
 - Conduction
 - Convection
 - Evaporation/condensation of water or steam
 - Radiation (not considered)



Heat transfer model

- Simplified model

Conduction/convection - Fourier

Geometry – plane sheet

No mass transfer phenomena considered

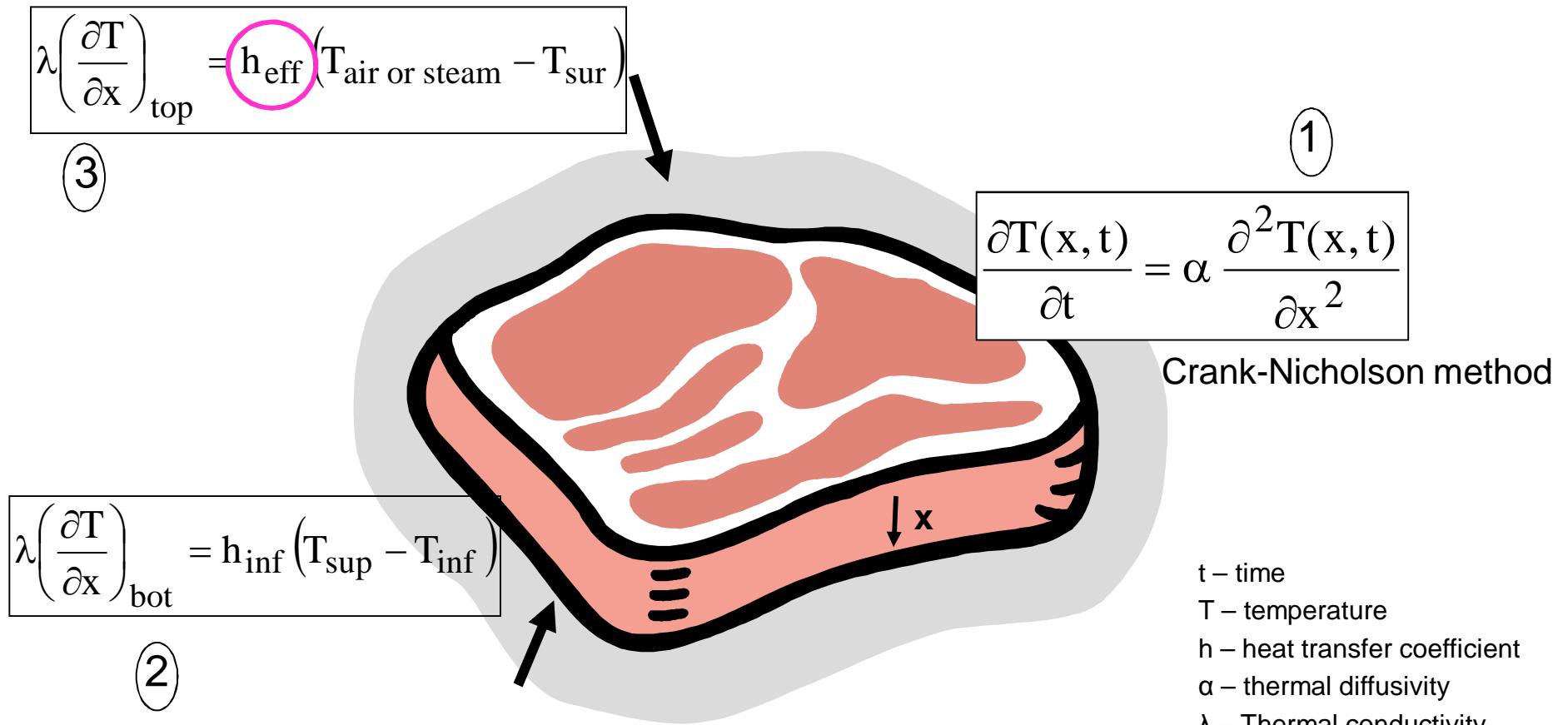


- Limits

Air velocity ranged from 15 m/s to 25 m/s

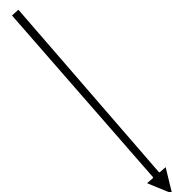
Extrapolation locked outside of conditions that can be verified in the test rig

Heat transfer model



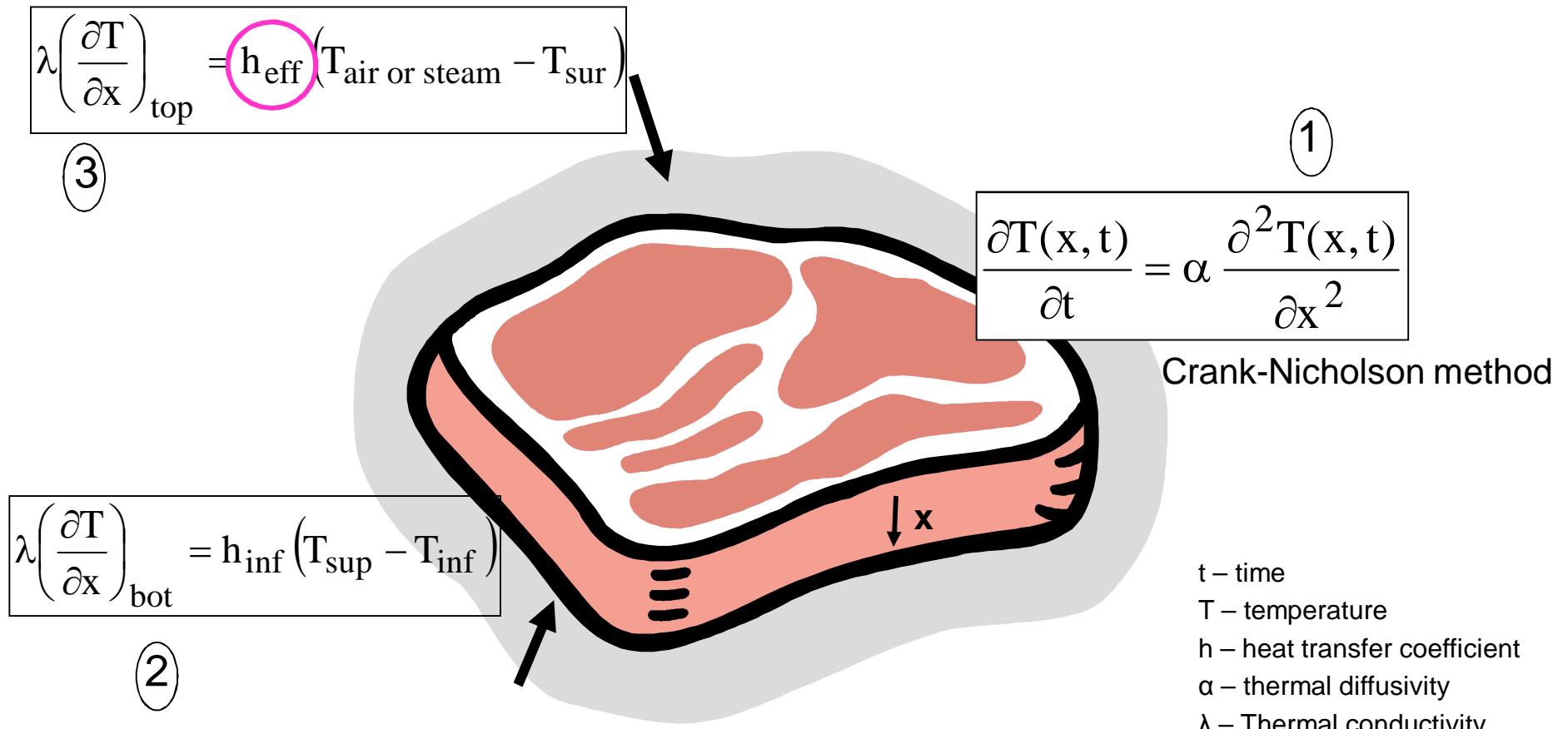
Boundary conditions - bottom

Exchanges between the bottom of the product and the support can be neglected – product thickness < 0.5 cm



Experimentally observed in the rig

Heat transfer model



t – time

T – temperature

h – heat transfer coefficient

α – thermal diffusivity

λ – Thermal conductivity

Effective coefficient – dry conditions

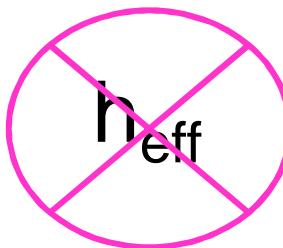
$$h_{\text{eff}} = h \frac{T_{\text{air}} - T_{\text{sur}}}{T_{\max} - T_{\text{sur}}} + K_m \Delta H \frac{P_T - a_w P_{\text{sur}}}{T_{\max} - T_{\text{sur}}}$$

convection evaporation

k_m – mass transfer coefficient
 ΔH – latent heat of water evaporation



Effective coefficient – wet conditions (steam)



Due to complexity of mathematical expressions to be incorporated in the software

$$T_{\text{sur}} = T_{\text{steam}} - 3^{\circ}\text{C}$$



Experimentally observed in the rig

Microbial inactivation model

- To predict microbial content at the surface of foods
- It has the advantage of dealing with time-varying temperature conditions
- Typical of pasteurisation treatments



Microbial inactivation model

$$\frac{dN}{dt} = -k_{\max} k_Q(Q) N$$

$$\frac{dQ}{dt} = -k_{\max} Q$$

N – microbial cell density

Q – variable related to the physiological state of the cells

k_{\max} – inactivation rate constant

$$\log \frac{N}{N_0} = \log(\exp(-k_{\max} t) \frac{1 + Q(0)}{1 + Q(0) \exp(-k_{\max} t)})$$

Microbial inactivation model

$$k_{\max}(T, a_w) = \frac{\ln 10}{D_{ref}} \exp\left(\frac{\ln 10}{z} (T - T_{ref})\right) \exp\left(\frac{\ln 10}{z_{a_w}} (a_w - 1)\right) + c1$$



$T = T_{sur}$ → Calculated on the basis of all considerations
of heat transport

Software Program



'Bugdeath' - funded by the European Commission under the EC Framework 5;
Quality of Life and Management of Living Resources Programme



Software Program

- The programme was developed using Real Basic® 5.2 application
- Food/microorganism selection is allowed (database of thermal properties and kinetic parameters)
- On the basis of the selection of a heating regime of the medium, the programme allows prediction of the food surface temperature and simulates the microbial load content along the whole process time



First screen – product/microorganism

BugDeath

Product Process Output

Product
Beef

Diffusivity: 1.23e-07 Conductivity: 4.5e-01

Product thickness (m): 0.016

Init. Product Temp. (°C): 13.4

Sample Diameter (m): 0.05

Water activity: 0.45

Microorganism
S. typhimurium

D (min): 6.6e+1 Safe Count: 1.e+0 Initial Count: 10000000

Z (°C): 7 Z(aw): 0.2 R. Temp. (°C): 57.2

C1: 0.2206 Q0: 2.279372077

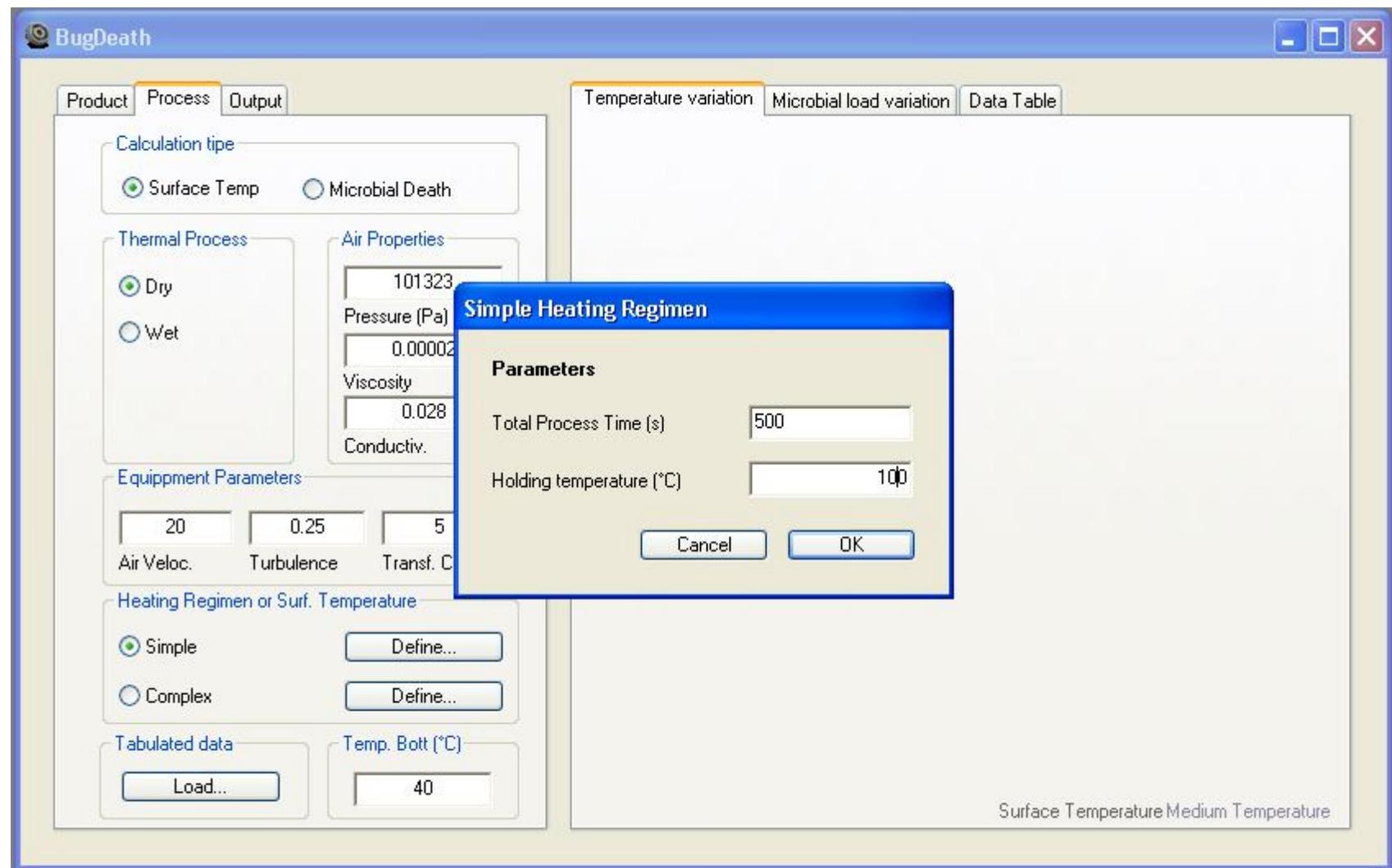
Temperature variation Microbial load variation Data Table

Surface Temperature Medium Temperature

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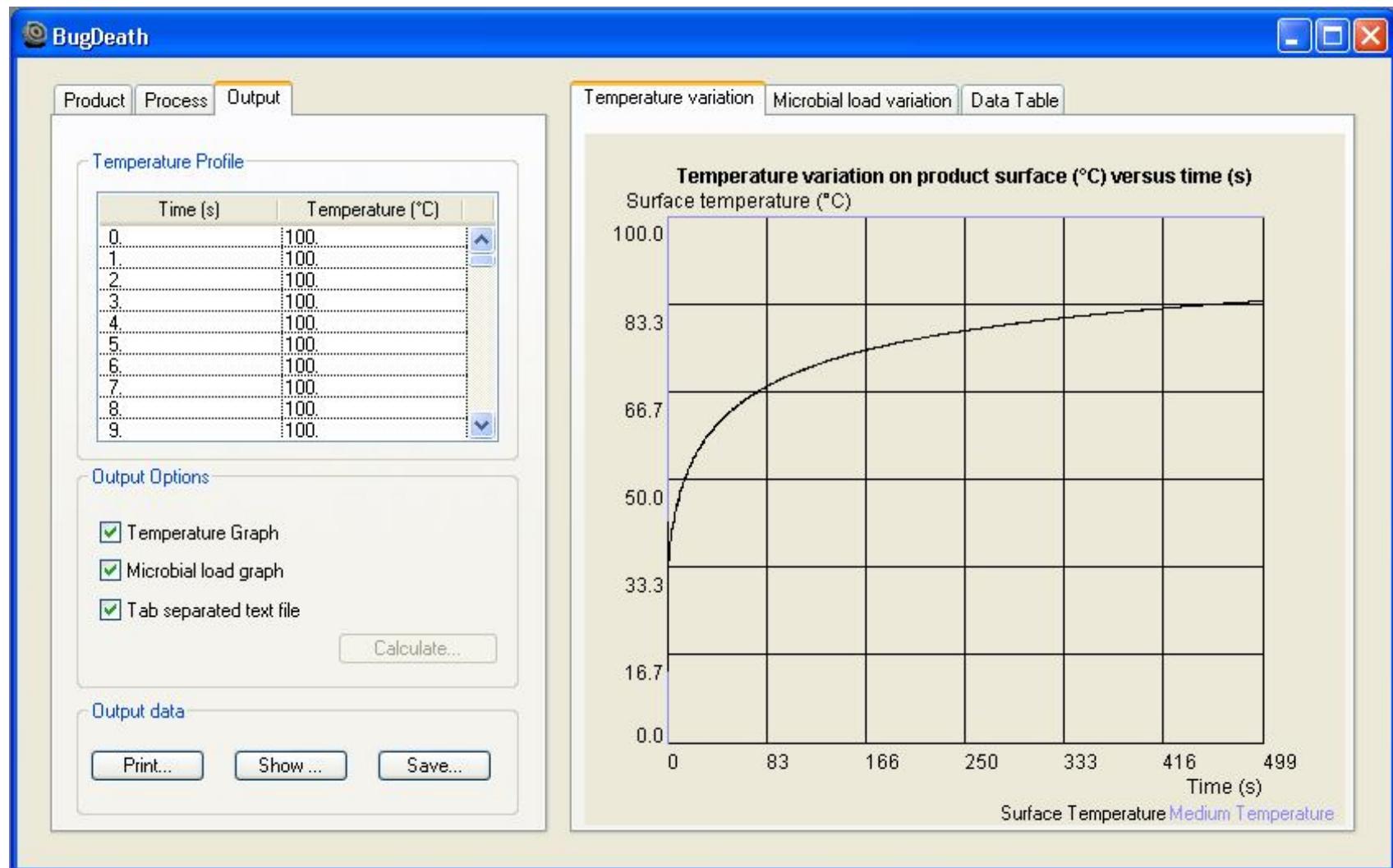
Second screen – process



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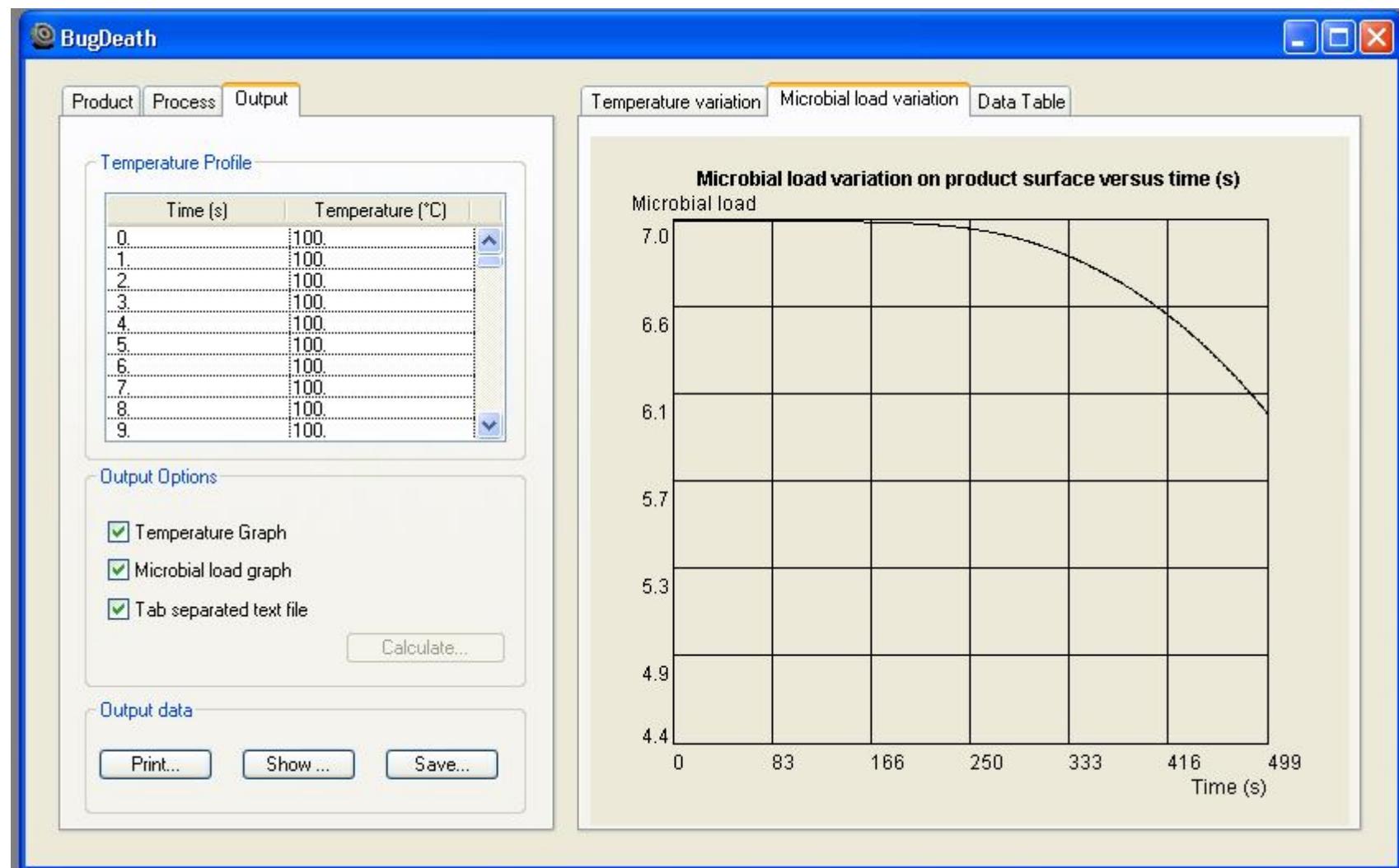
Output – graphic/temperature



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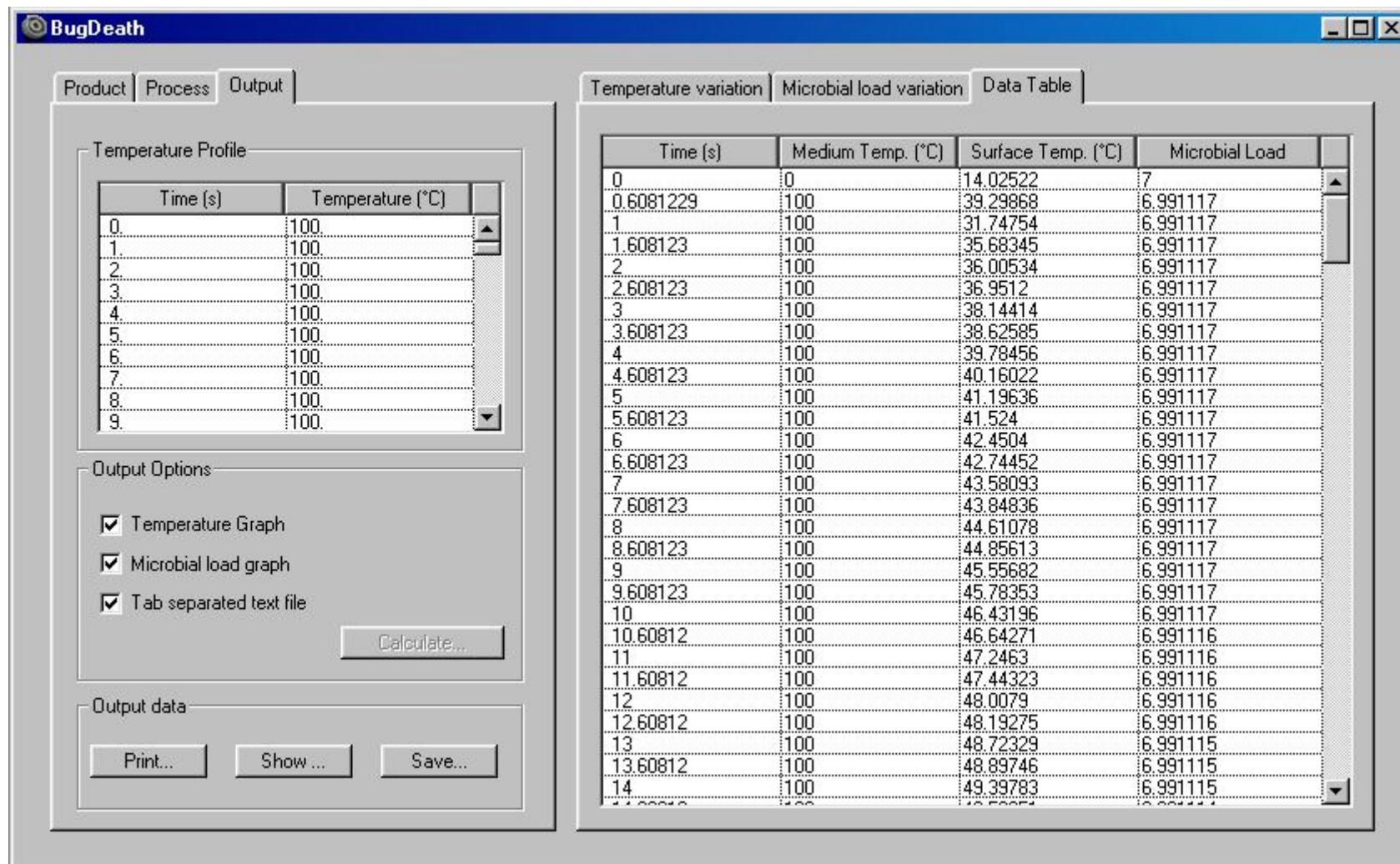
Output – graphic/microbial load



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Output – data table



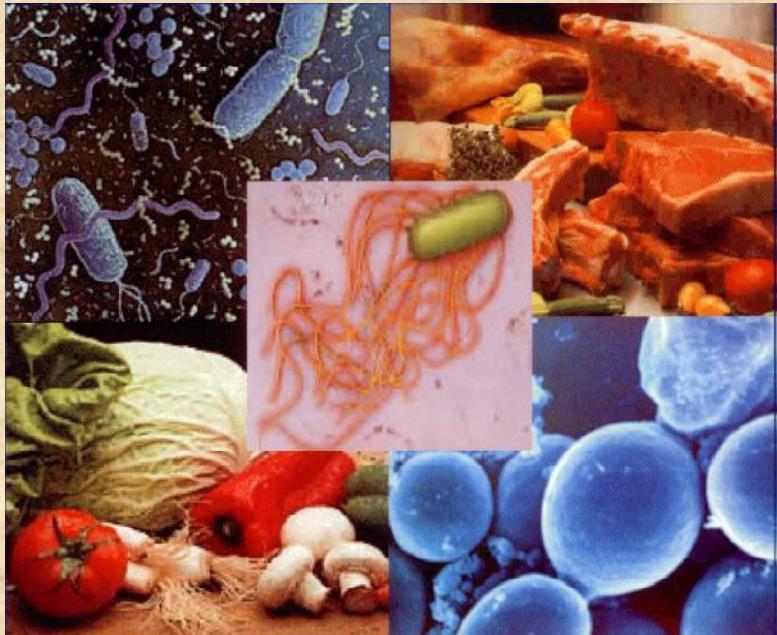
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Conclusions / outputs

- Software application simulates the results obtained in the rig apparatus
- Valuable for developing appropriate and safety thermal processes
- Marketed and commercially available
- Educational purposes (simulation of real food processes)





Predictive Microbiology

A tool to support food safety decisions

Thank you

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Escola Superior de Biotecnologia

7th December 2007