



Modelling Food Quality Changes Kinetics During Thermal Processing

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Outline

- Objectives of thermal processing
- Importance of thermal processes
- Factors affecting quality changes
- Modelling approaches
- Quality changes kinetics
- Predictive quality
- Combining with other technologies
- Post-harvest treatment
- Challenges

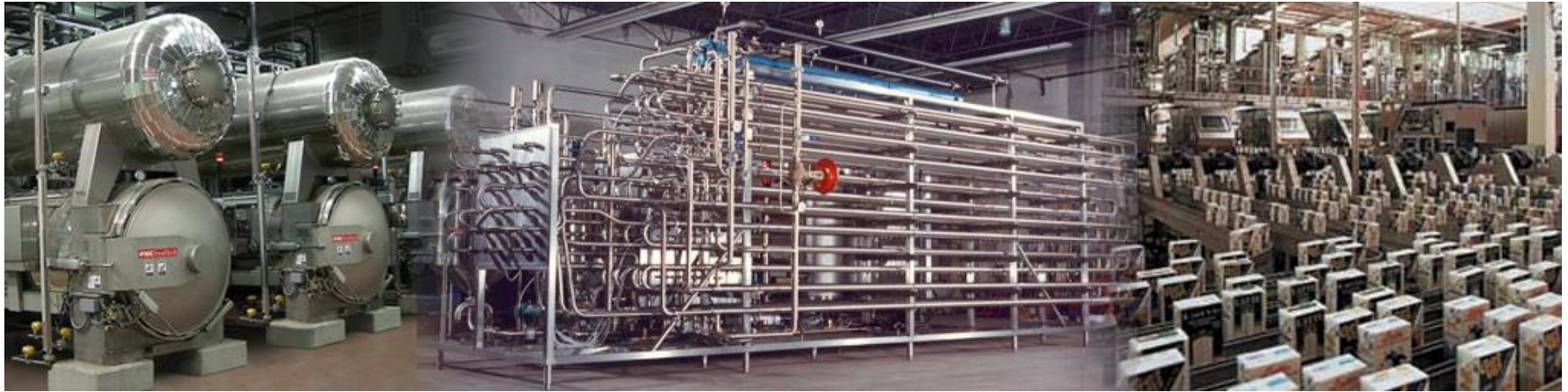


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Objectives of thermal processing



... Originally designed to inactivate

spoiling and pathogenic microorganisms

and enzymes



***bacteria
yeasts
molds***



Objectives of thermal processing

... consumers request

- ✓ Fresh like, appealing, convenient, healthy

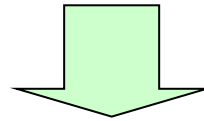


- ✓ Environmental care



Objectives of thermal processing

Thermal Processes



Should

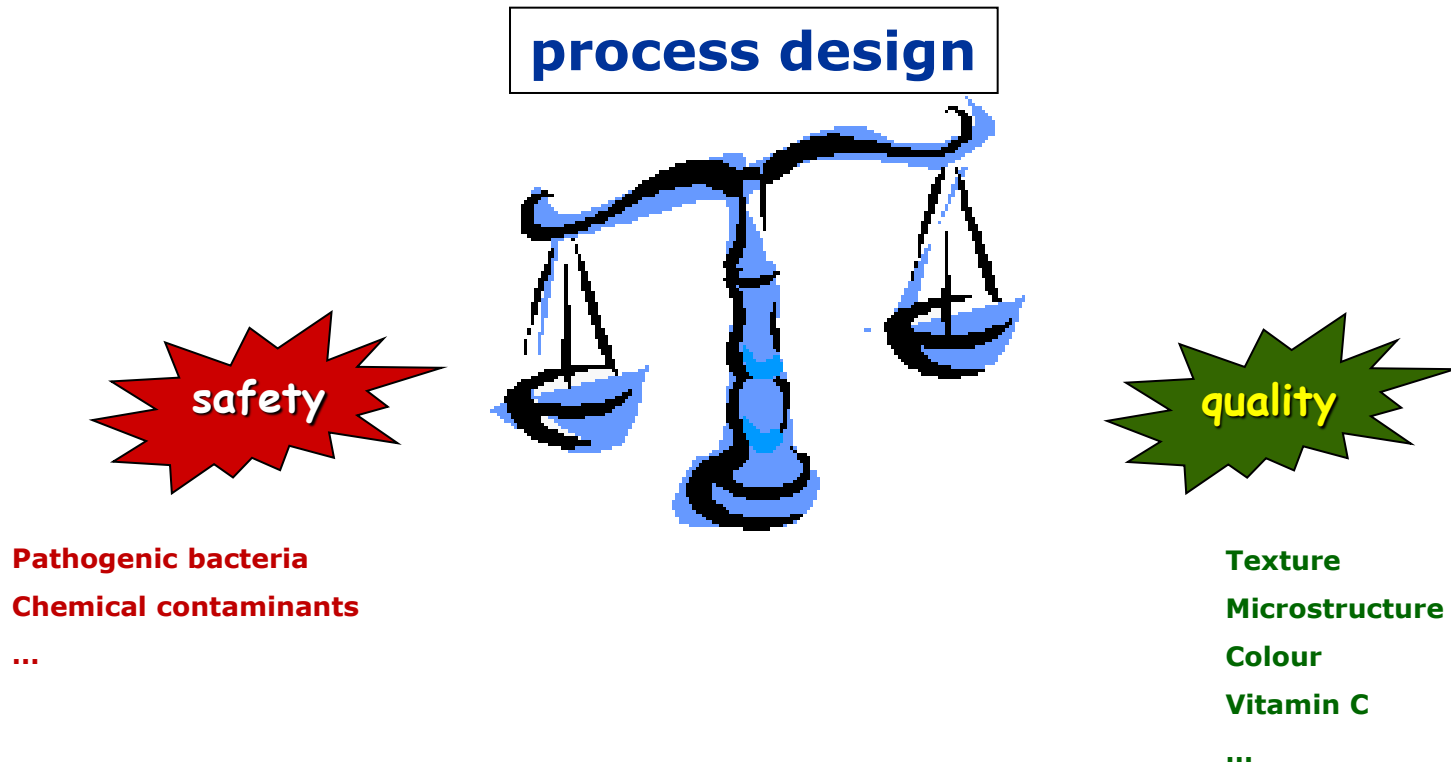
prevent the degradation of the original **organoleptic** and **nutritive** food characteristics

quality



Objectives of thermal processing

thermal processes affect negatively quality factors



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Importance of thermal processes

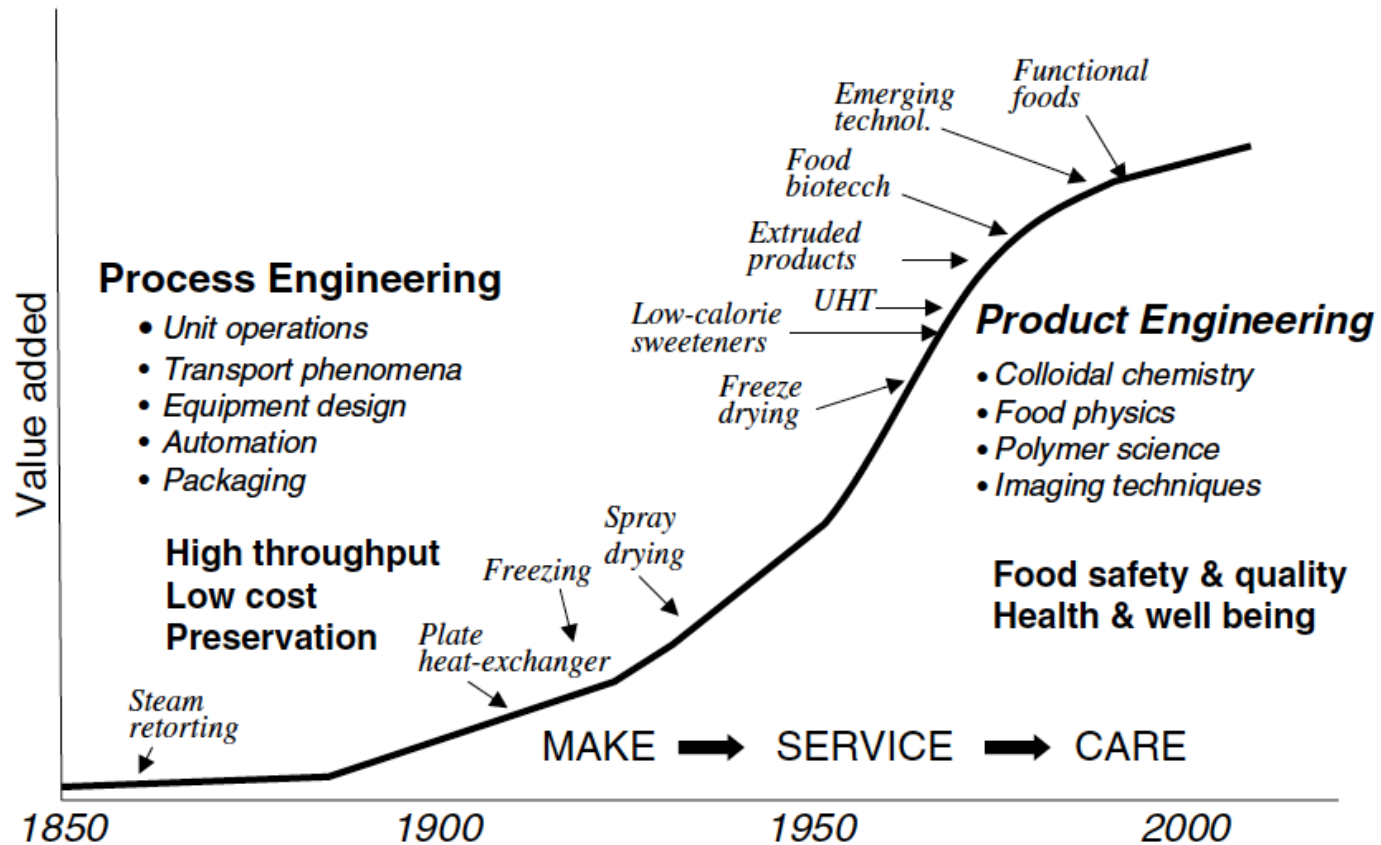


Figure 1. Evolution of the food industry in terms of value added to products and shift in emphasis from process engineering to product engineering. This transition has implied a change in concepts and techniques that support each approach.

¹Aguilera J. (2006). *J. Sci. Food & Agric.* 86(8): 1147-55.



Importance of thermal processes

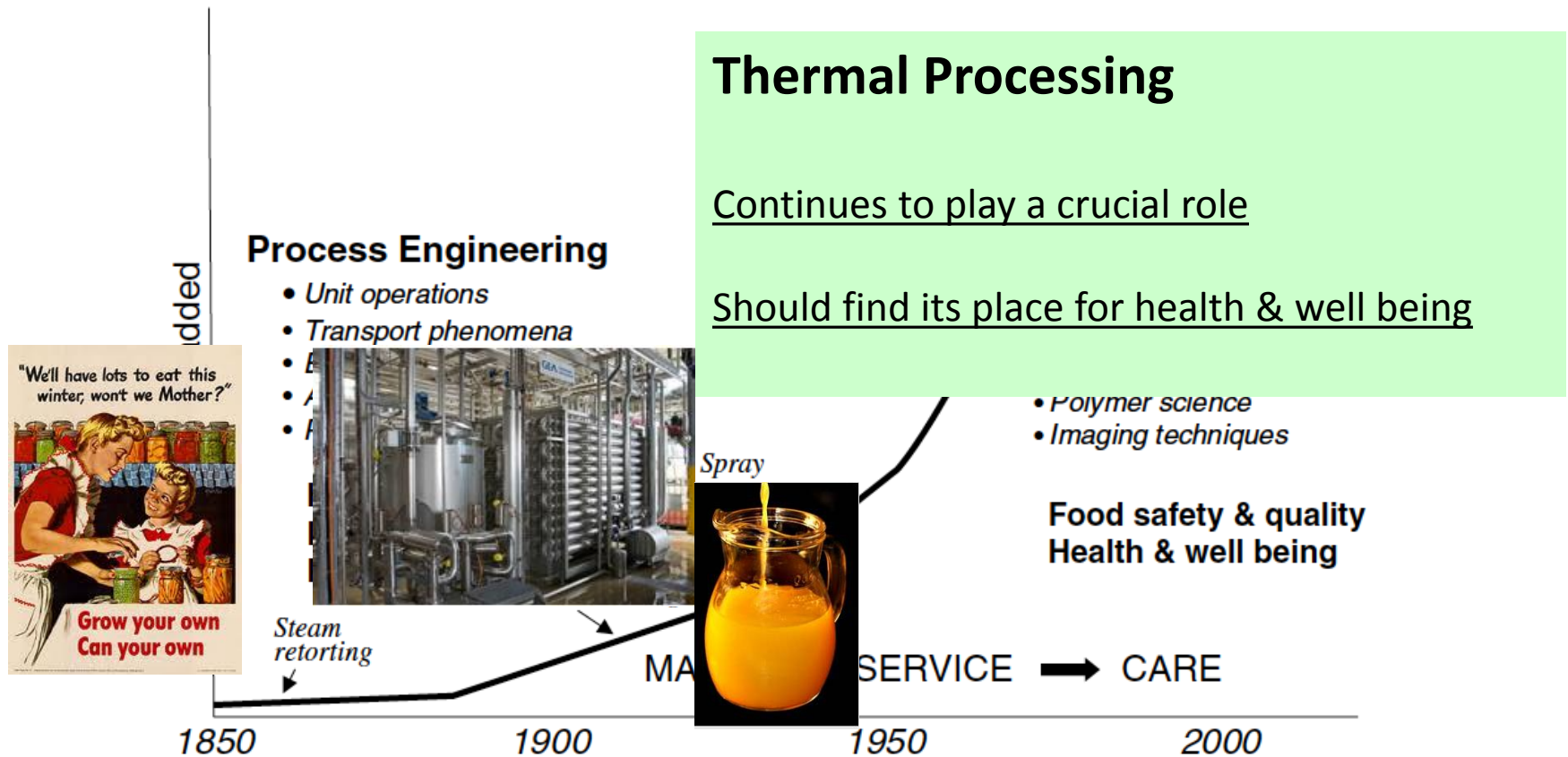
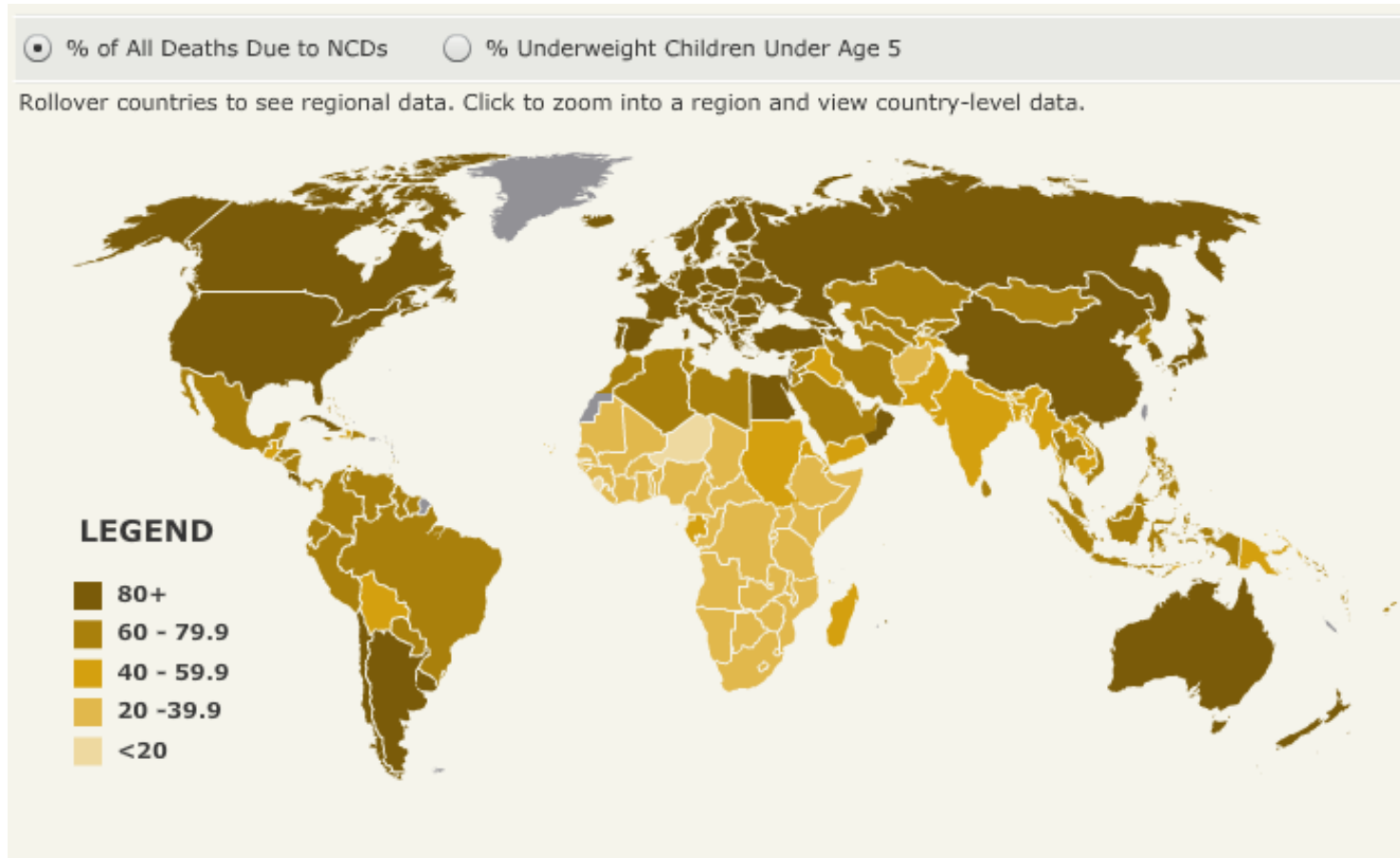


Figure 1. Evolution of the food industry in terms of value added to products and shift in emphasis from process engineering to product engineering. This transition has implied a change in concepts and techniques that support each approach.

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Importance of thermal processes



population ageing



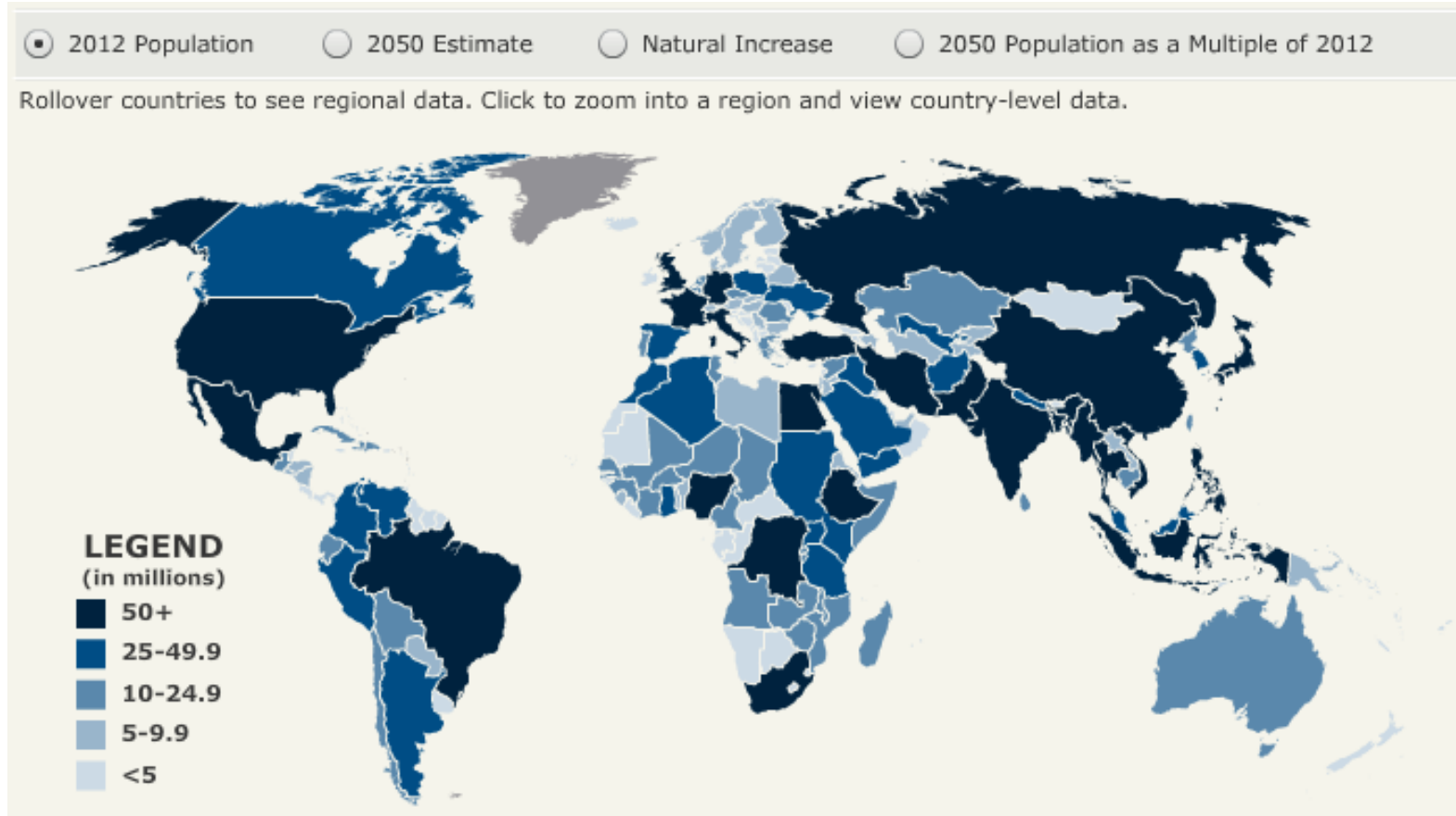
Importance of thermal processes



population ageing



Importance of thermal processes



population growth



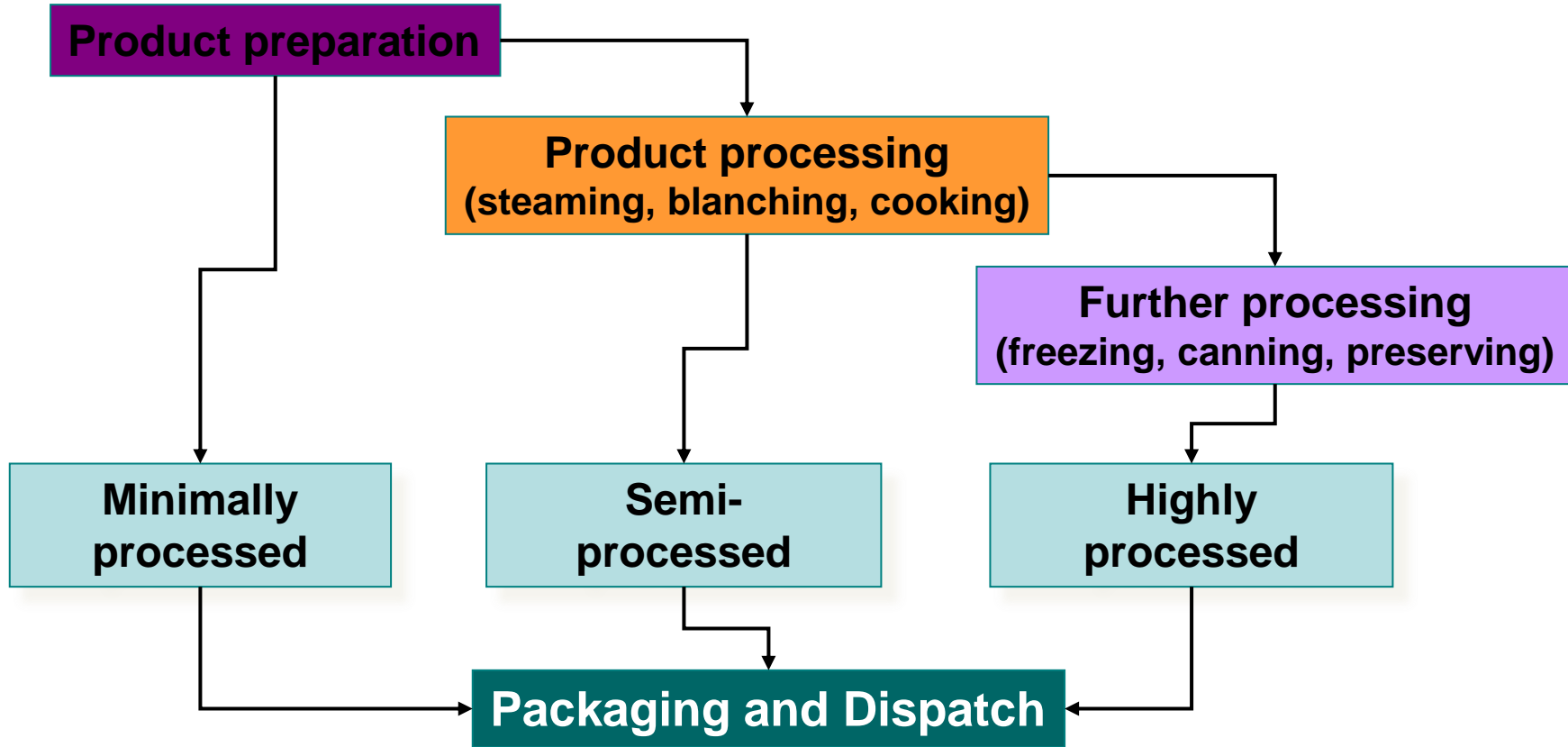
Importance of thermal processes



population growth



Importance of thermal processes



Outline

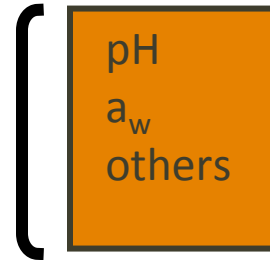
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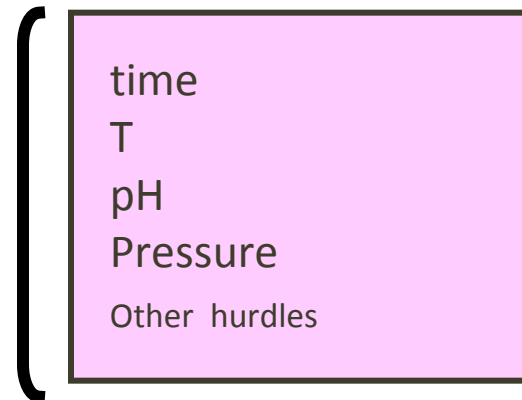
Factors affecting quality changes

Quality attributes response depends on:

- **Intrinsic factors**



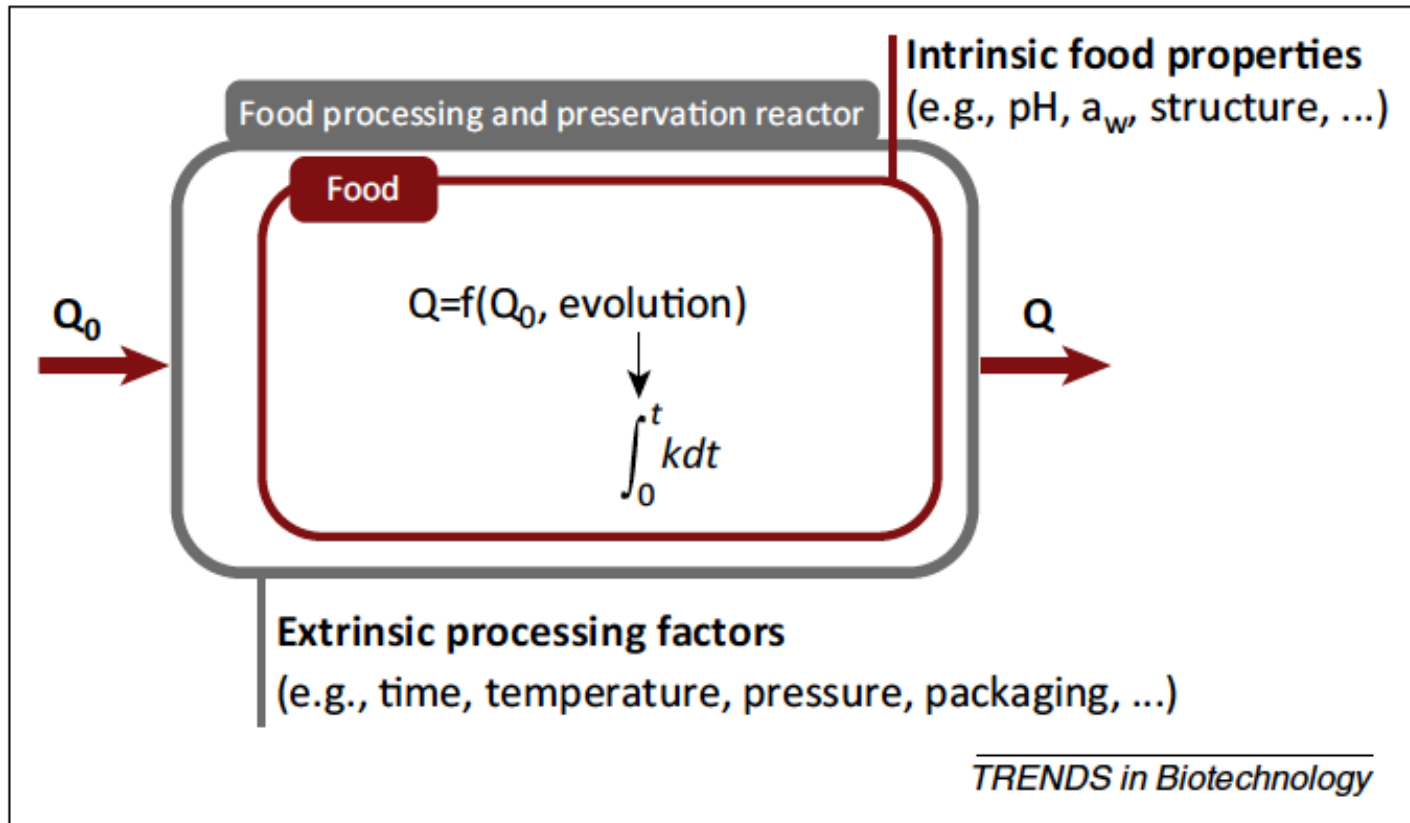
- **Extrinsic factors**



- **System dynamics**



Factors affecting quality changes



Grauwet et al, 2014



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Modelling approaches

Model → mathematical expression

$$y_i = f(x_{ij}, \theta_k) + \varepsilon_i$$

$i=1,2,\dots,n$ (number of experimental runs/observations)

$j=1,2,\dots,v$

$k=1,2,\dots,p$

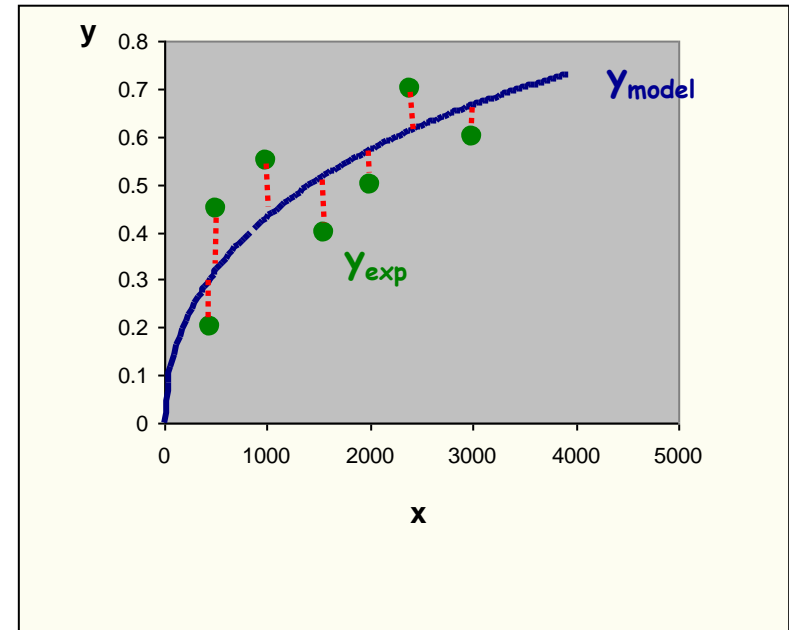
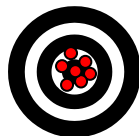
Minimize differences



Precise ?

θ^*

Accurate ?



Modelling approaches

objective

precise and accurate description of observations

model adequacy

quality of model parameters



Modelling approaches

Sampling:

- Heuristic sampling
- Experimental design

Minimize variance of:

- *predicted response*
- *parameter estimates*



Modelling approaches

Data analysis:

- Regression schemes

$$SSR = \sum_{i=1}^n e_i^2 = \sum_{i=1}^n [y_i - f(x_{ij}, \theta_k)]^2$$

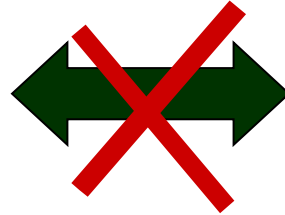
Least-squares
method

- Analysis of residuals



Modelling approaches

Mathematical complexity



Adequate description

quality

model



parameters



Modelling approaches

advantages

- **knowledge of the process**
- **process effects on product**
- **control of process variables**



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Quality changes kinetics

Historically: → single-response studies

→ static conditions



Quality changes kinetics

Single-response → empirical

Kinetic Models

Zero order

$$C = C_0 - k_{(T)}t$$

1st order

$$C = C_0 e^{-k_{(T)}t}$$

Fractionary

$$\frac{C - C_{eq}}{C_0 - C_{eq}} = e^{-kt}$$

Equilibrium value

Attribute at time t

Initial value

Time

Activation energy

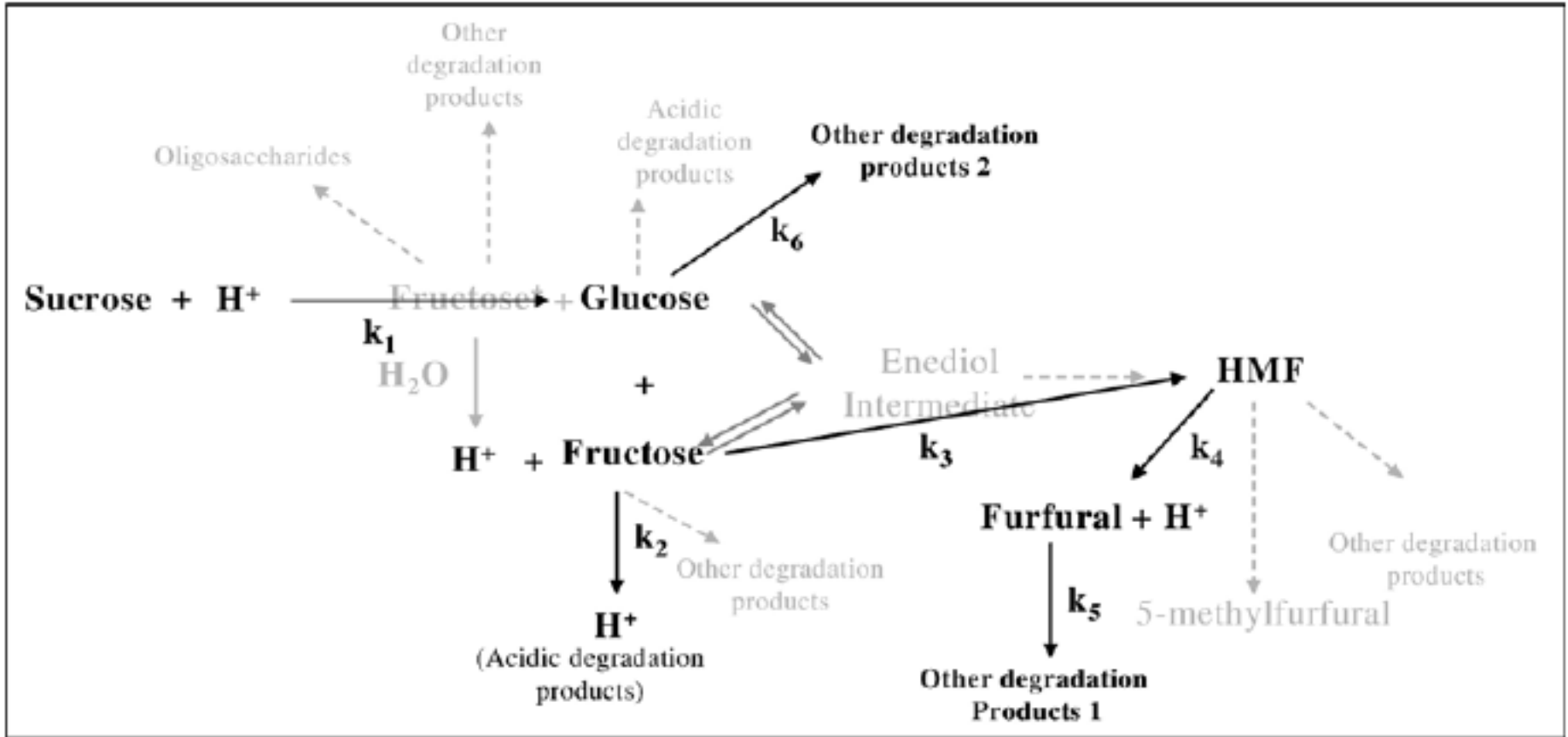
$$k_{(T)} = k_{ref} \exp \left[\frac{E_a}{R} \left(\frac{1}{T} - \frac{1}{T_{ref}} \right) \right]$$

T effect on k
Arrhenius law



Quality changes kinetics

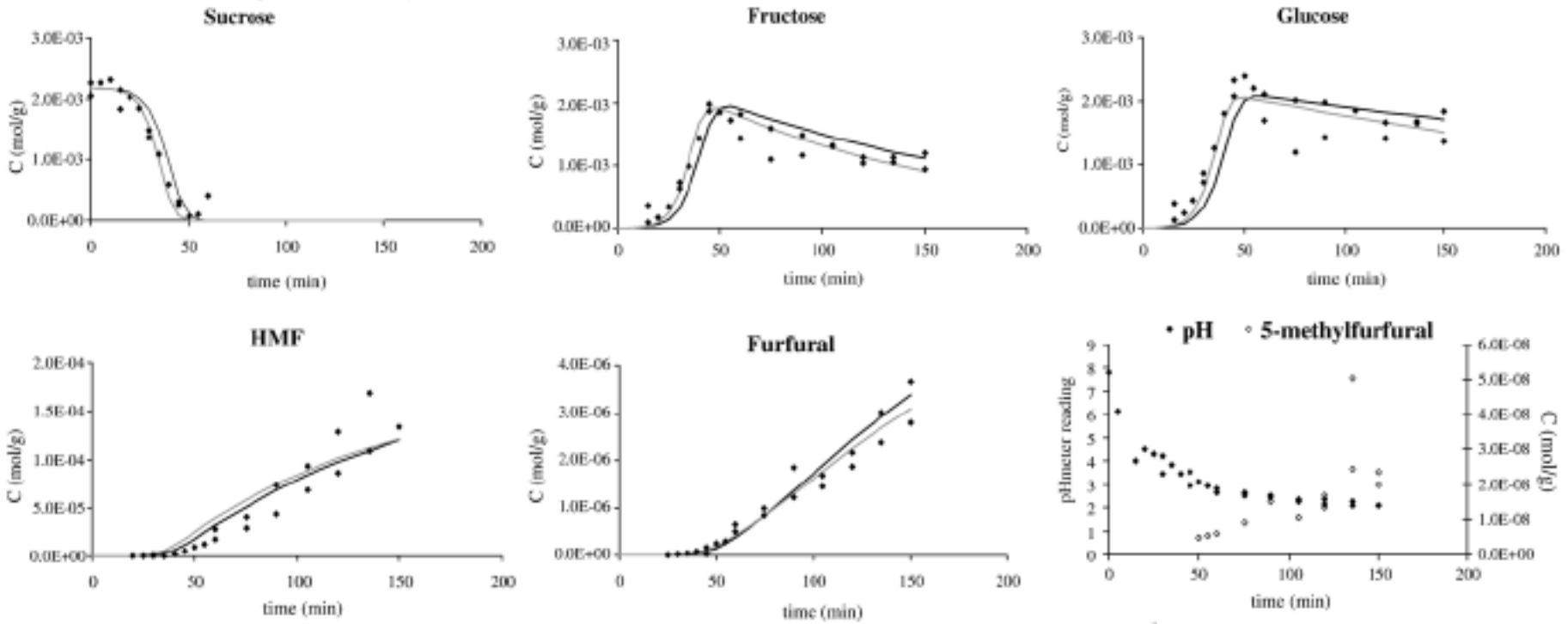
Multiresponse Modelling → mechanistic



Quintas et al, 2007



Quality changes kinetics



Quintas et al, 2007



Quality changes kinetics

Laboratory research



Industrial scale



Quality changes kinetics

The complexity of dynamic conditions

Laboratory research

- Studies are often carried out at constant temperatures

Industrial scale

- Time-varying temperature conditions are common

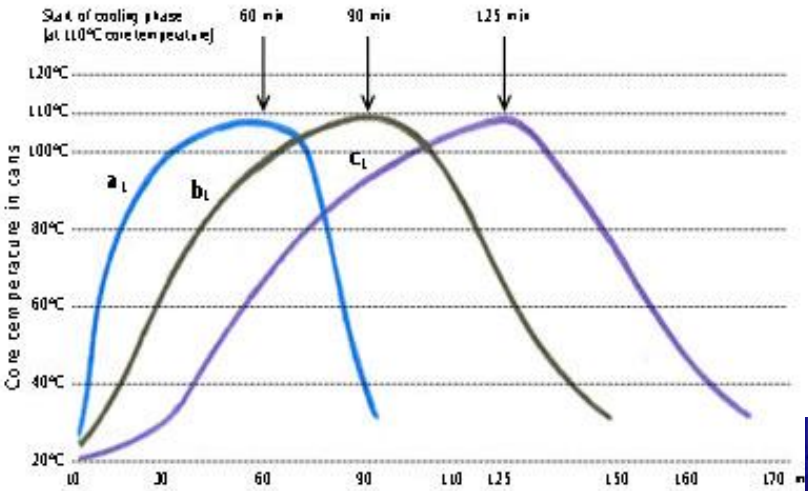
transfer of results is compromised

isothermal

non-isothermal



Quality changes kinetics



→ Need to have temperature histories and distributions



Quality changes kinetics



→ Need to have temperature histories and distributions

→ Hot - filling



Quality changes kinetics

Isothermal conditions

$$C = C_0 - \left(k_{\text{ref}} \exp \left(- \frac{E_a}{R} \left(\frac{1}{T} - \frac{1}{T_{\text{ref}}} \right) \right) t \right)$$

$$C = C_0 \exp \left(- k_{\text{ref}} \exp \left(- \frac{E_a}{R} \left(\frac{1}{T} - \frac{1}{T_{\text{ref}}} \right) \right) t \right)$$

$$C = C_{\text{eq}} + (C_0 - C_{\text{eq}}) \exp \left(- k_{\text{ref}} \exp \left(- \frac{E_a}{R} \left(\frac{1}{T} - \frac{1}{T_{\text{ref}}} \right) \right) t \right)$$

Non-Isothermal conditions

$$C = C_0 - \left[k_{\text{ref}} \int_0^t \exp \left[- \frac{E_a}{R} \left(\frac{1}{T(t)} - \frac{1}{T_{\text{ref}}} \right) \right] dt \right]$$

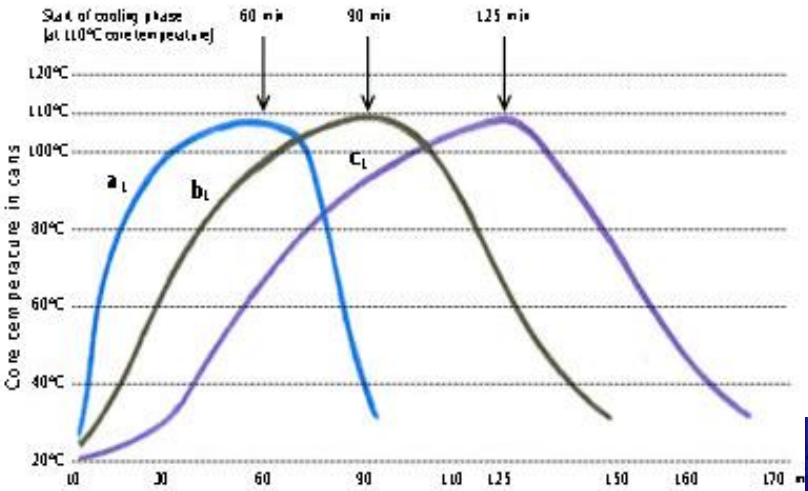
$$C = C_0 \exp \left[- k_{\text{ref}} \int_0^t \exp \left[- \frac{E_a}{R} \left(\frac{1}{T(t)} - \frac{1}{T_{\text{ref}}} \right) \right] dt \right]$$

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Integration



Quality changes kinetics



→ Reverse Engineering can be an excellent tool



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Predictive quality



What's NEW?

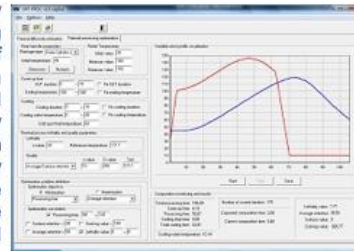
25/10/2010
OPT-PROx 1.0 alpha-version
is released.

12/01/2012
OPT-PROx 1.5 alpha-version
is released.

REVIEWS

...manufacture of shelf stable
canned foods, and has been
the cornerstone of the food
processing industry for more
than a century" (A. Teixeira)

"OPT-PROx" is a software specially developed for thermal food processing numerical optimization. The diversity of thermal food processing optimization problems with different objectives and required constraints are solvable by OPT-PROx software. The adaptive random search algorithm coupled with penalty functions approach, and the finite difference method with cubic spline approximation are utilized by OPT-PROx for simulation and optimization of thermal food processes. The possibility of numerical estimating the thermal diffusivity coefficient based on the mean squared error function minimization is included. "OPT-PROx" software was successfully tested on the real thermal food processing problems.



The following objective functions and constraints are supported by "OPT-PROx" software.

Objective functions:

- Minimization of total processing time.
- Minimization of cooking value.
- Maximization of Surface quality retention.
- Maximization of Average quality retention.

Constraints:

- Surface quality retention.
- Average quality retention.
- Cooking value.
- Thermal lethality value.
- Total thermal processing time.



Predictive quality

Predictive microbiology

The use of **mathematical models** in the description of **microbial responses** to environmental stressing factors



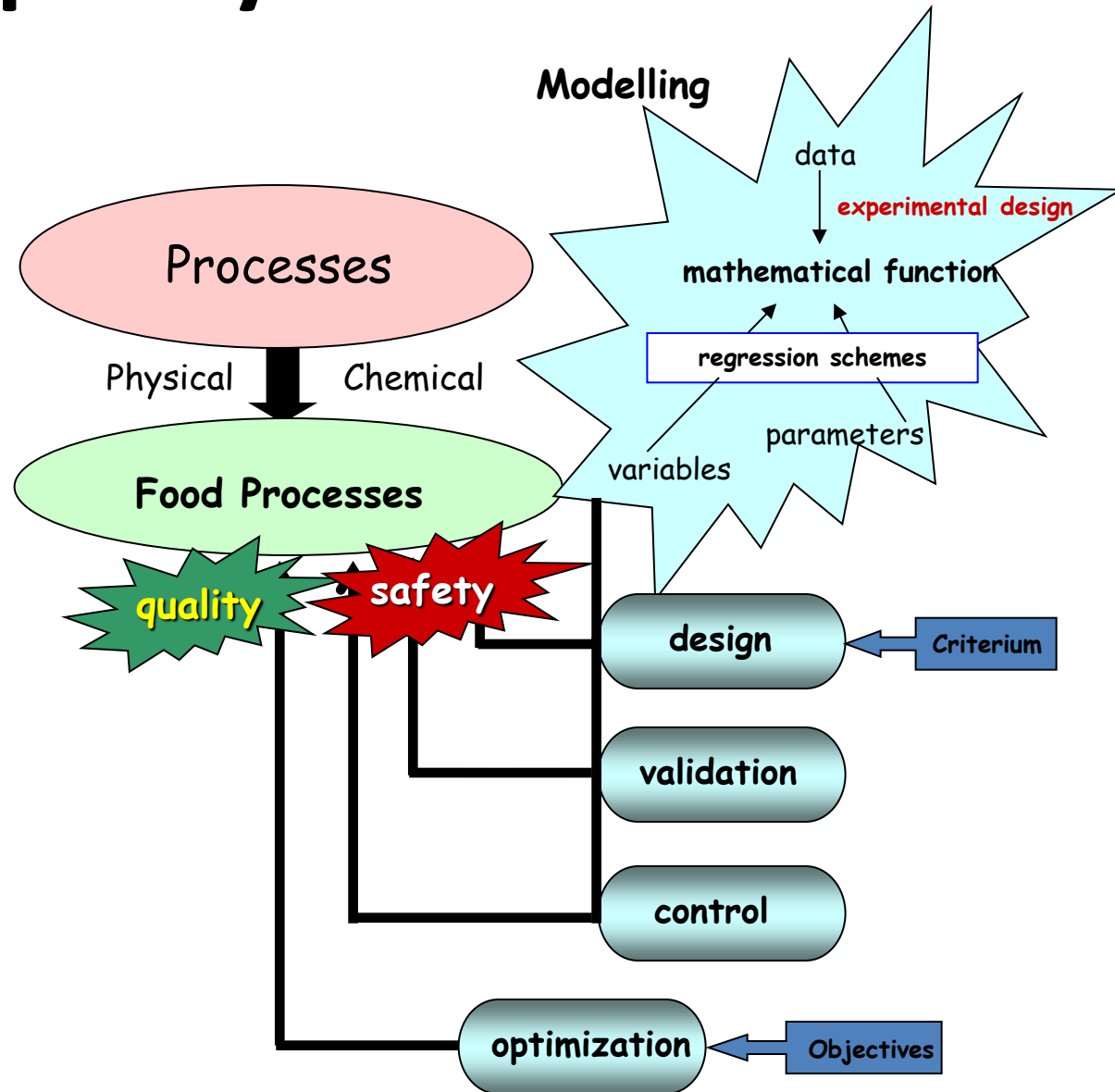
Predictive quality

Transport Phenomena

- heat
- mass
- *momentum*

Reaction kinetics

Properties

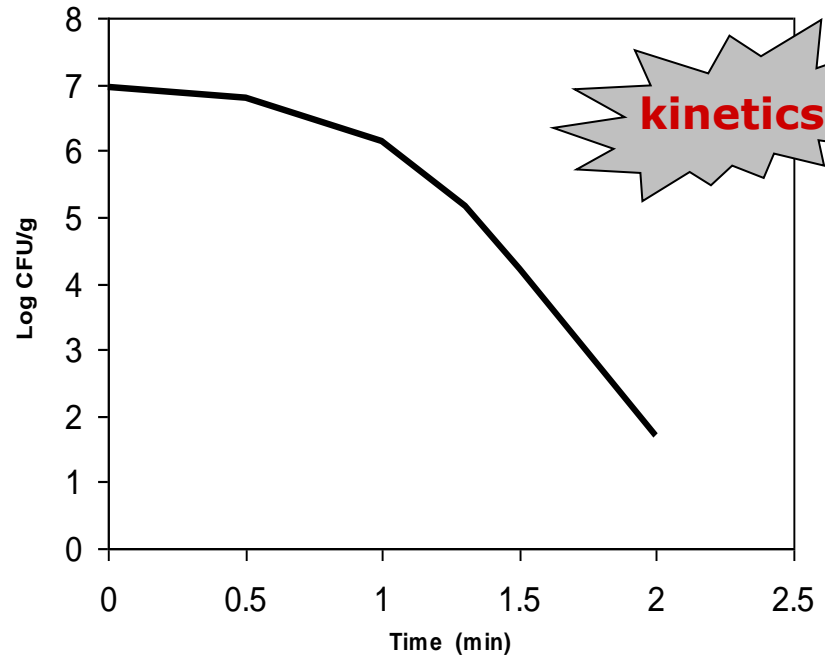


Predictive quality

❖ primary model



parameters

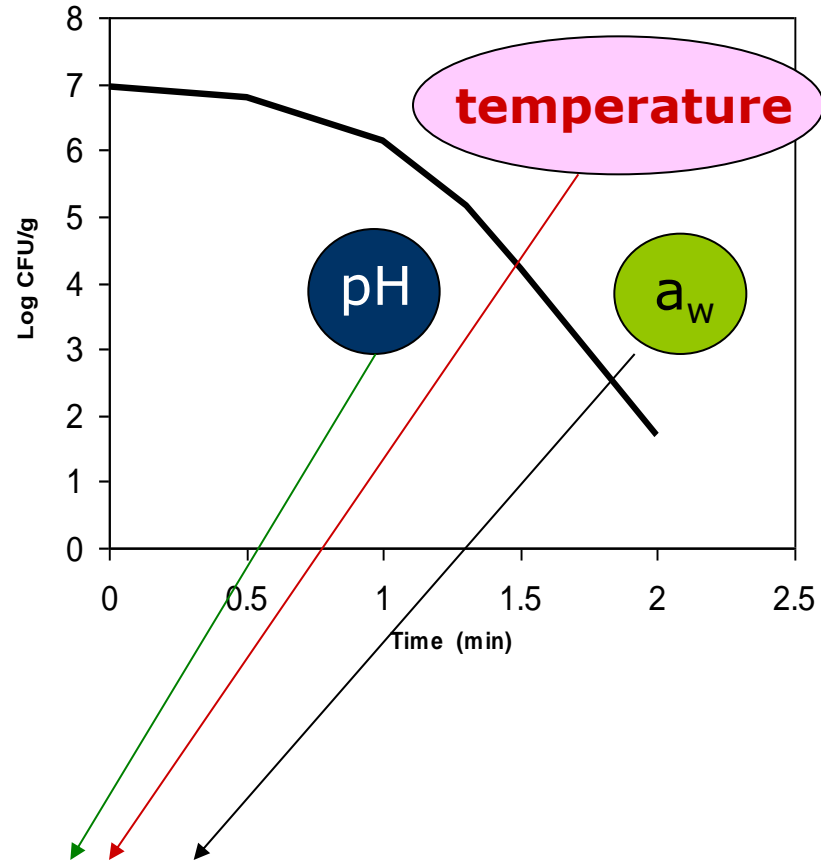


Predictive quality

❖ **primary model**

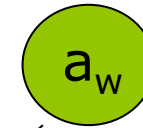
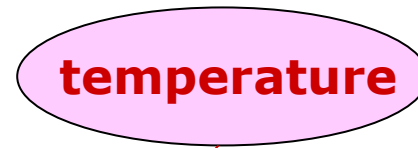
❖ **secondary model**

parameters



Predictive quality

❖ **primary**



❖ **secondary**

parameters

❖ **terciary** - integration of the previous models - **software**



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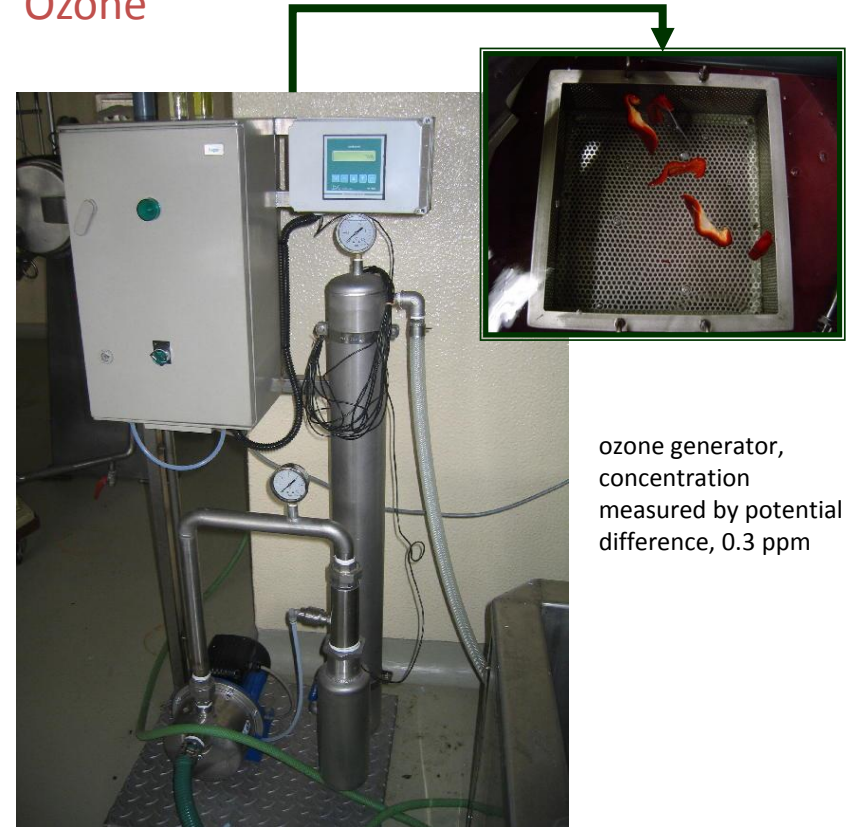
Combining with other technologies

UV-C radiation

UV-C chamber (University of Algarve), 4 germicidal UV lamps (TUV G30T8, 16 W, Philips, peak emission at 254 nm), average intensity 12.36 W/m²



Ozone



ozone generator, concentration measured by potential difference, 0.3 ppm

Ultrasonication / Thermosonication



ultrasound equipment (Bandelin Sonorex RK 100H) operating at 32 kHz

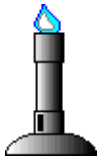
Thermal processing conference, *Campden BRI, Chipping Campden, Gloucestershire – UK, 12-13 June, 2014*



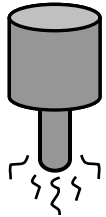
Combining with other technologies



Types of combined treatments with ultrasound



+



Heat + Ultrasound

Thermosonication



Combining with other technologies



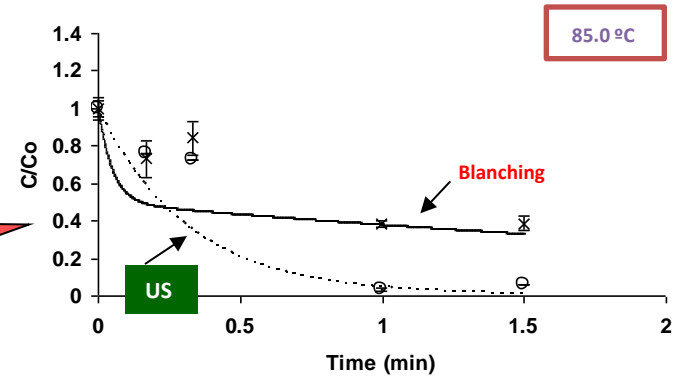
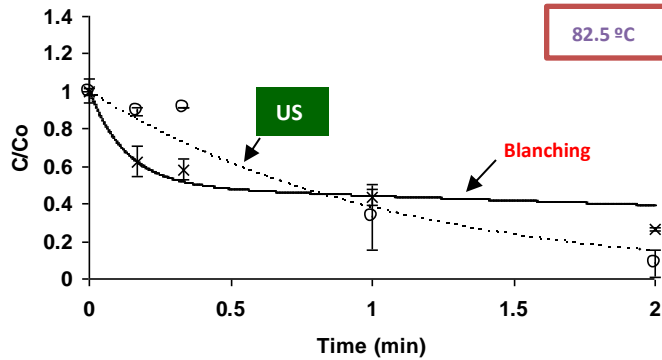
Colour changes

Vitamin C

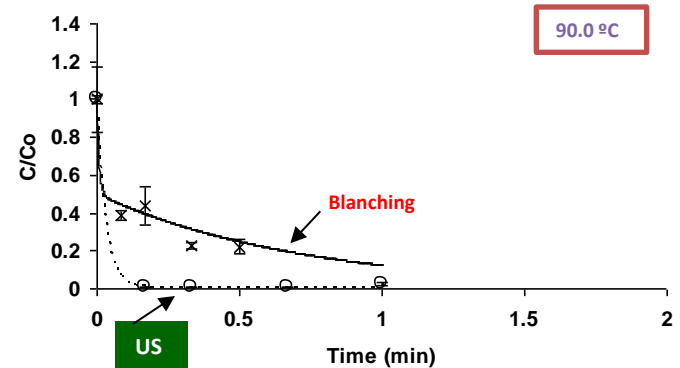
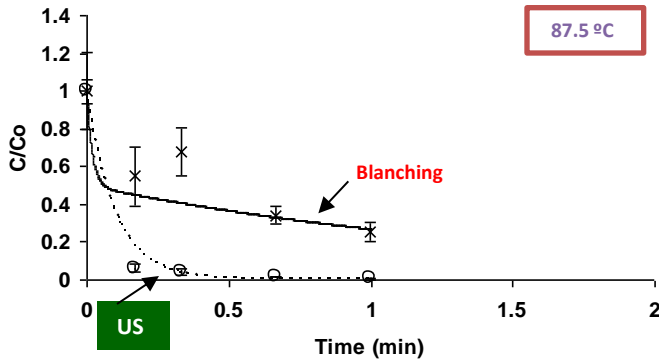
Peroxidase



Combining with other technologies



Peroxidase



Cruz et al, 2014



Combining with other technologies

The application of thermosonication



- temperatures above 85 °C and for the same blanching times

led to higher enzyme inactivation when compared to heat blanching processes

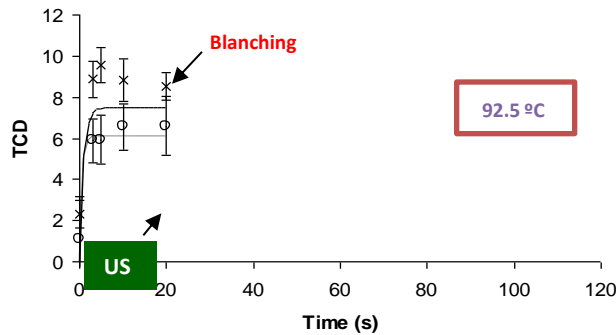
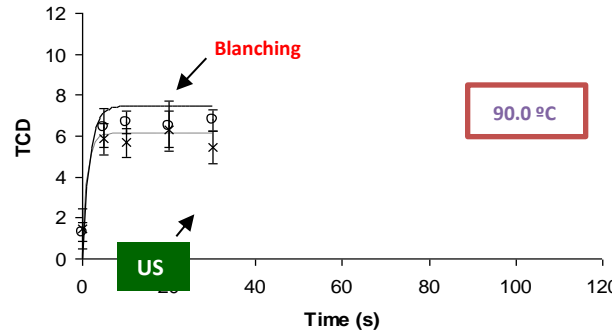
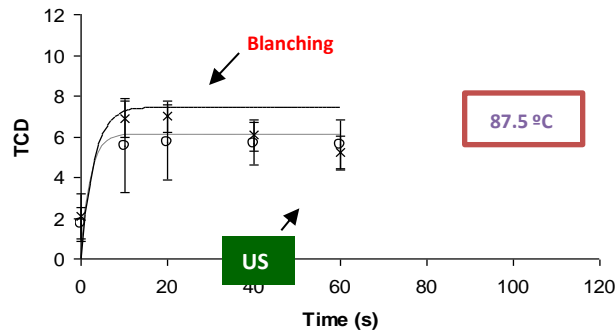
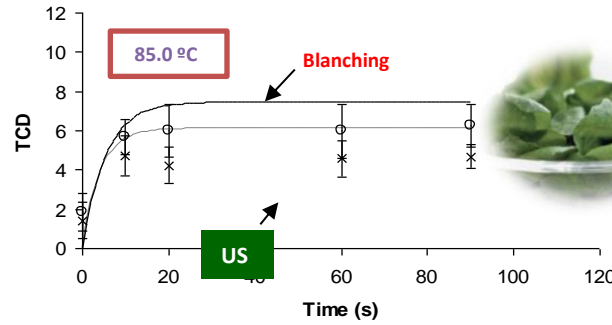
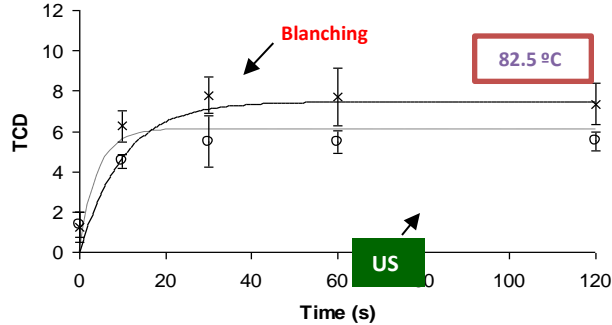
Peroxidase

These results allow the application of shorter blanching times at this range of temperatures, leading to a product with a higher quality, or minimized processing



Combining with other technologies

85 °C



Cruz et al, 2014

Combining with other technologies

The application of thermosonication

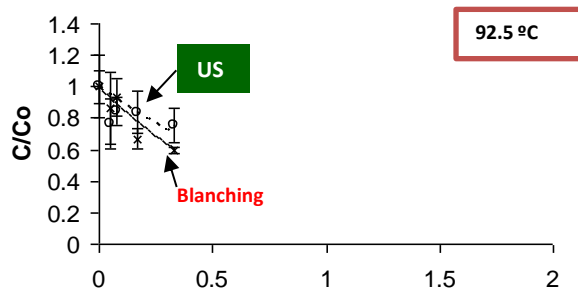
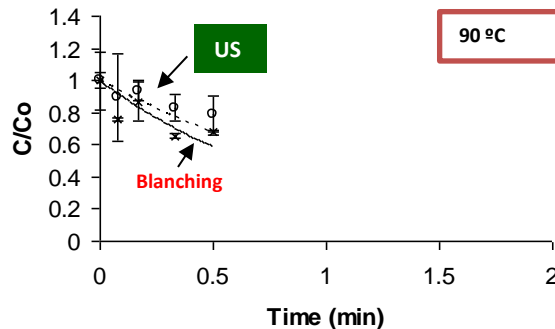
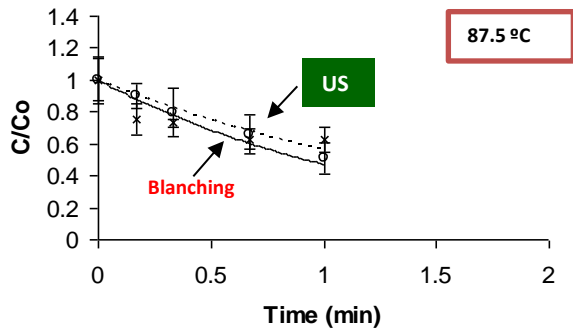
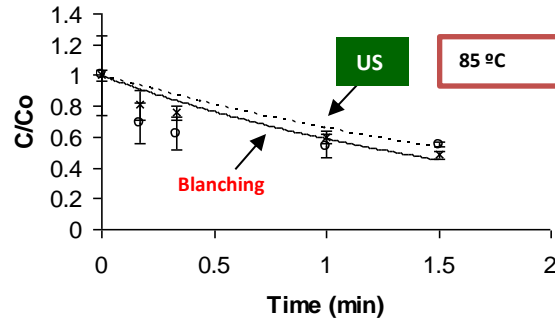
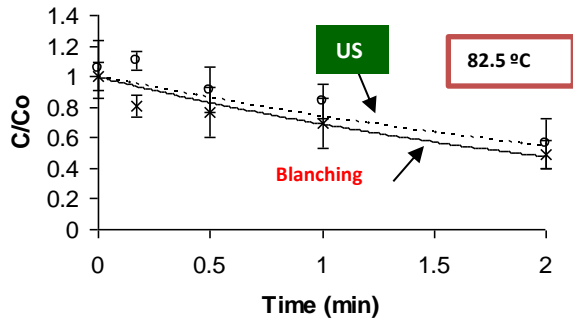


Colour

Reaction rates of watercress colour changes due to heat and thermosonication blanchings were not significantly different



Combining with other technologies



Cruz et al, 2014



Combining with other technologies

The application of thermosonication



Vitamin C

Results showed no significant differences between heat and thermosonication treatments

The treatment will allow good vitamin C retention



Combining with other technologies

The application of thermosonication



Quality

The thermosonication treatments can be a good alternative to the traditional heat blanching processes, since higher quality products are attained



Outline

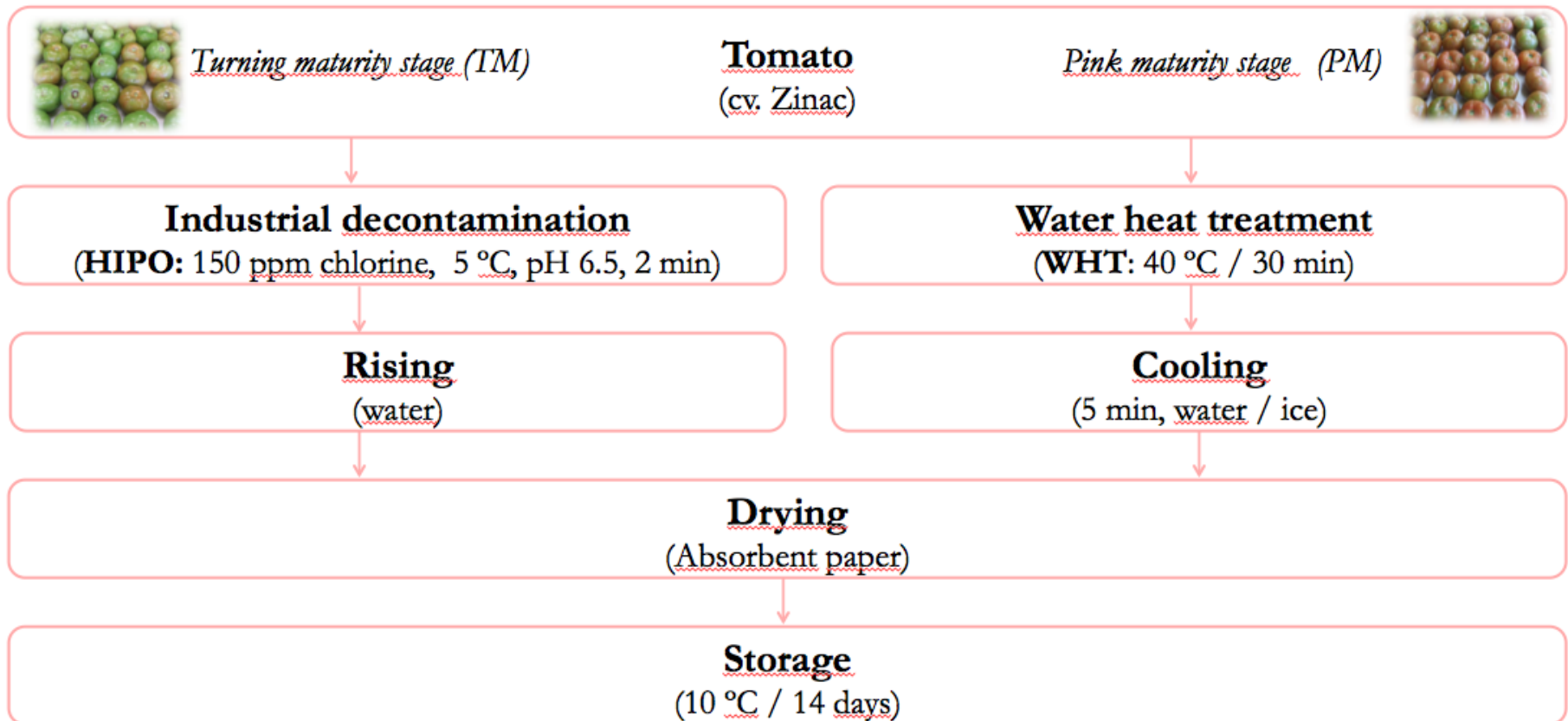
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Postharvest treatment



Postharvest treatment

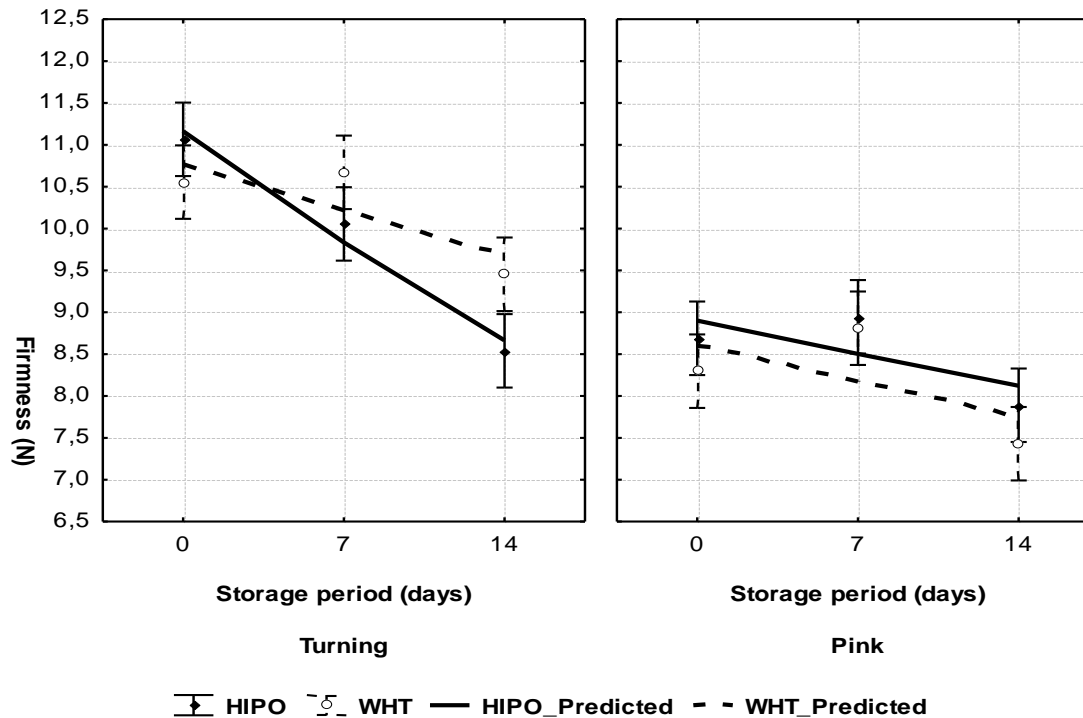


Pinheiro et al, 2014



Postharvest treatment

Firmness

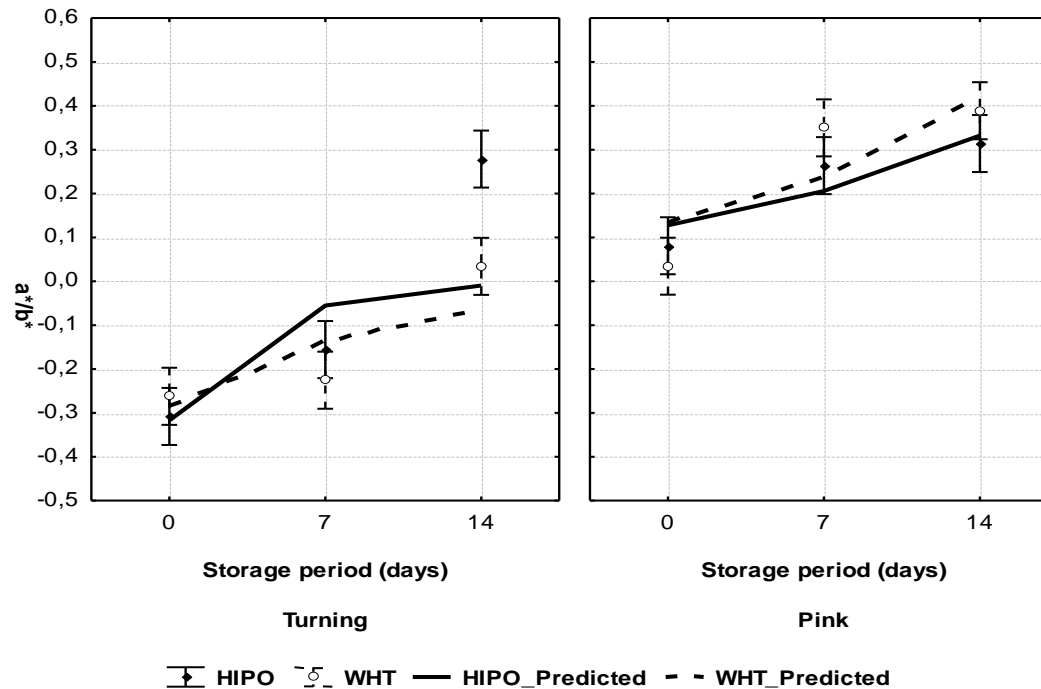


Pinheiro et al, 2014



Postharvest treatment

Colour



Pinheiro et al, 2014



Postharvest treatment

$$C = C_0 e^{-k_{(T)}t}$$

Maturity stages	Treatment	a^*/b^*	Firmness (N)
Turning	HIPO	$C_0 = -0.32 \pm 0.12$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.25 \pm 0.32$	$C_0 = 11.16 \pm 0.52$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.02 \pm 0.006$
	WHT	$C_0 = -0.28 \pm 0.11$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.11 \pm 0.10$	$C_0 = 10.77 \pm 0.69$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.01 \pm 0.01$
Pink	HIPO	$C_0 = 0.13 \pm 0.07$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = -0.07 \pm 0.05$	$C_0 = 8.90 \pm 0.43$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.01 \pm 0.01$
	WHT	$C_0 = 0.13 \pm 0.07$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = -0.08 \pm 0.04$	$C_0 = 8.60 \pm 0.72$ $k_{10^\circ\text{C}} (\text{day}^{-1}) = 0.01 \pm 0.01$

→ Results provide strong evidence that postharvest water **heat treatment (40 °C - 30 min)** for tomato fruits (cv. 'Zinac') at turning maturity stage guarantees the overall quality at 10 °C, **twice as long of fruits washed with chlorinated water.**



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Challenges

- ✓ Food quality kinetic studies under dynamic conditions.
- ✓ Use of reverse engineering.
- ✓ Kinetic studies for combined processes.
- ✓ Multiresponse models – mechanistic.
- ✓ More intelligent processing – better process control, design and optimization.
- ✓ Development of a so called “**Predictive Quality**”.





Thank you

Rui Cruz

Mafalda Quintas

Teresa Brandão

Elsa Gonçalves

Joaquina Pinheiro

