

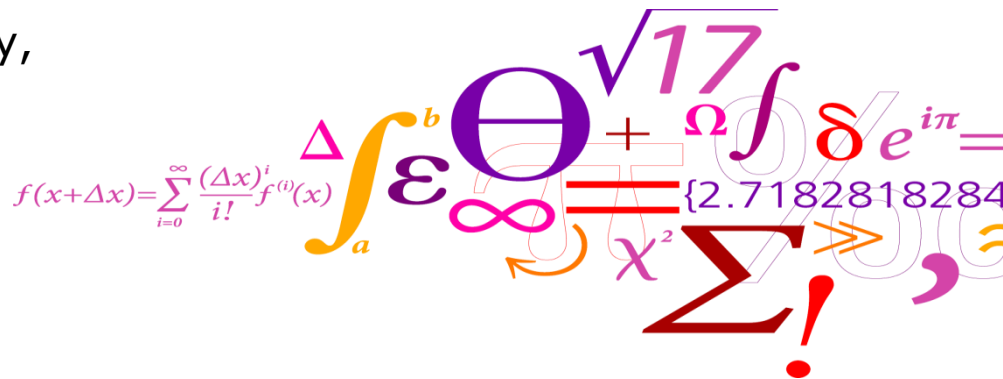


Predictive microbiology for the dairy industry

Veronica Martinez-Rios, Ioulia Koukou, Marie Jørgensen, Sarah Kadhim,
Paw Dalgaard

Analytical and Predictive Microbiology,
National Food Institute (DTU Food),
Technical University of Denmark

DTU Food
National Food Institute

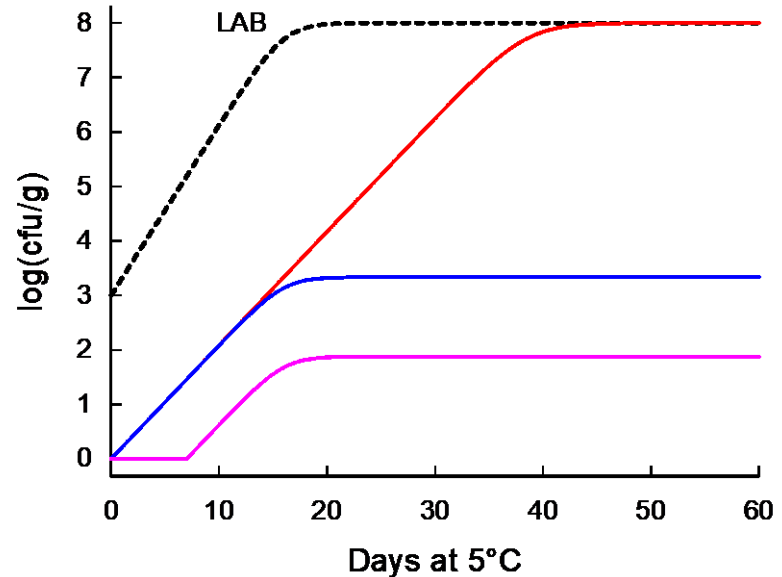
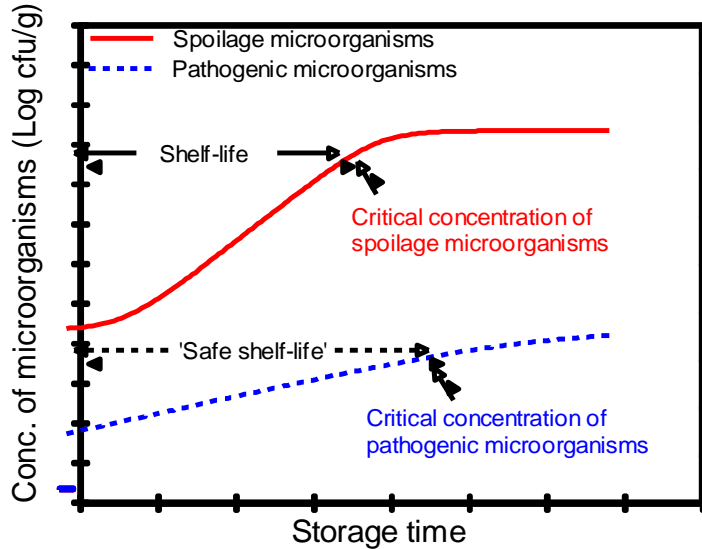


Predictive microbiology for the dairy industry

Outline

- Application of predictive food microbiology
- Evaluation of models and software
- New cardinal parameters and evaluation
- Examples of application for dairy products
- Conclusions and perspectives

Predictive microbiology - concept



Processing
Product characteristics
Storage conditions



Safe shelf-life
Optimal product recipes
Spoilage and shelf-life

Product characteristics

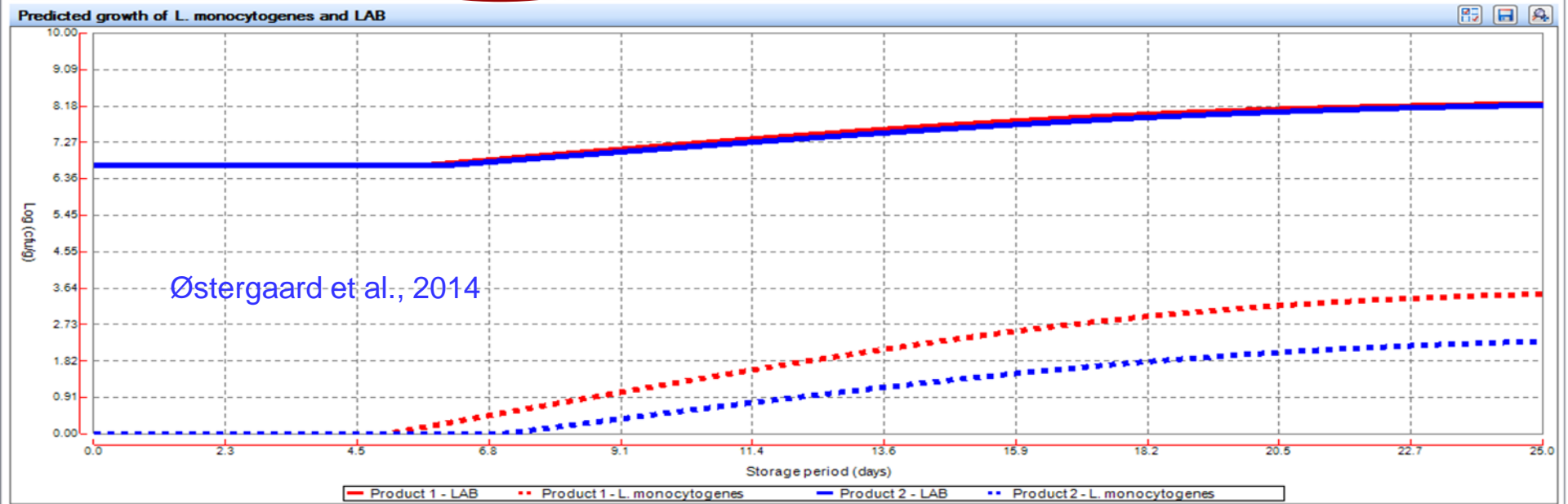
Listeria monocytogenes, cfu/g Lactic acid bacteria, cfu/g LAB aroma culture used Worst case relative lag time Storage period (d)	Product 1	Product 2	Temperature (°C) Salt in water phase of product, % Initial pH Lactic acid in water phase of product, mg/l Sorbic acid in waterphase of product, mg/l	Product 1	Product 2
	1	1		7.0	7.0
	5000000	5000000		1.0	1.0
	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>		5.47	5.20
	<input type="checkbox"/>	<input type="checkbox"/>		450	1700
	25			0	0

Apply Clear



Constant temperature Series of constant temperatures Temperature profiles from logger data

Listeria monocytogenes			Lactic acid bacteria			
	Growth rate μ_{max} (1/h)	Lag phase (d)	Time for 100-fold increase (d)	Growth rate μ_{max} (1/h)	Lag phase (d)	Time to reach 8.0 log cfu/g (d)
Product 1	0.0257	5.06	13.13	0.0119	5.69	18.55
Product 2	0.0184	7.05	20.08	0.0111	6.07	19.8



Product characteristics

	Product 1	Product 2		Product 1	Product 2
<i>Listeria monocytogenes</i> , cfu/g	1	1	Temperature (°C)	7.0	7.0
Lactic acid bacteria, cfu/g	5000000	5000000	Salt in water phase of product, %	1.0	1.0
LAB aroma culture used	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	Initial pH	5.47	5.47
Worst case relative lag time	<input type="checkbox"/>	<input type="checkbox"/>	Lactic acid in water phase of product, mg/l	450	450
Storage period (d)	25		Sorbic acid in water phase of product, mg/l	0	750

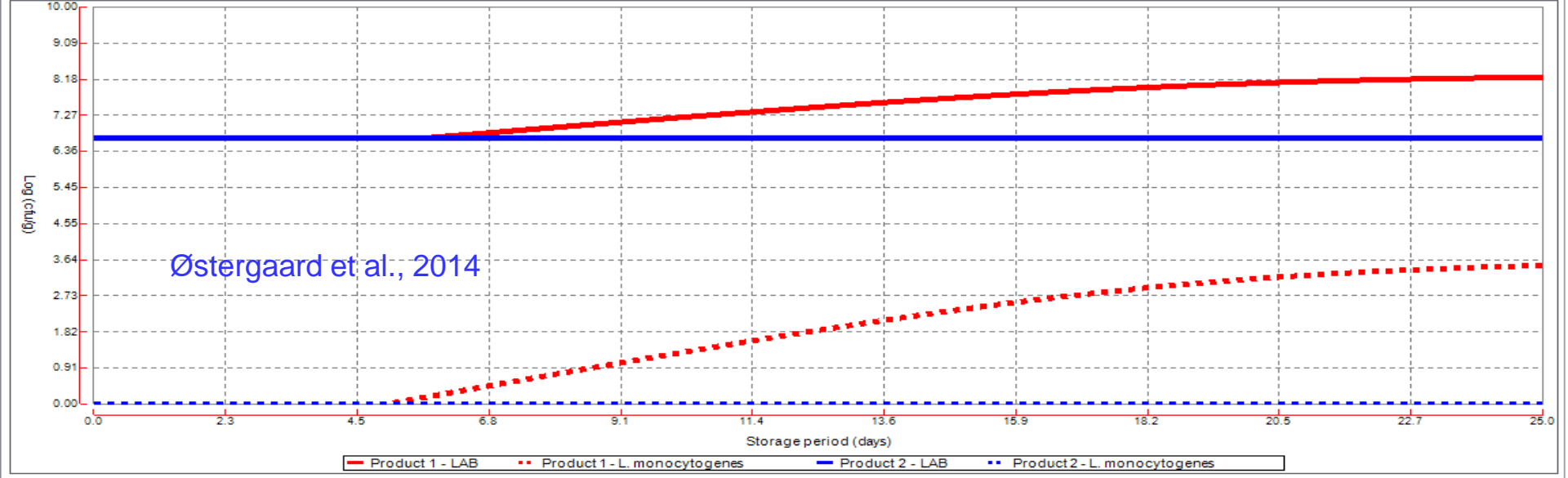
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Constant temperature Series of constant temperatures Temperature profiles from logger data

	<i>Listeria monocytogenes</i>			Lactic acid bacteria		
	Growth rate μ_{max} (1/h)	Lag phase (d)	Time for 100-fold increase (d)	Growth rate μ_{max} (1/h)	Lag phase (d)	Time to reach 8.0 log cfu/g (d)
Product 1	0.0257	5.06	13.13	0.0119	5.69	18.55
Product 2	0.0047	27.64	Not reached	0	Above 40 d.	Not reached

Predicted growth of *L. monocytogenes* and LAB





Growth of *Listeria monocytogenes* in cottage cheese in combination with LAB (starter or aroma culture)

Product characteristics

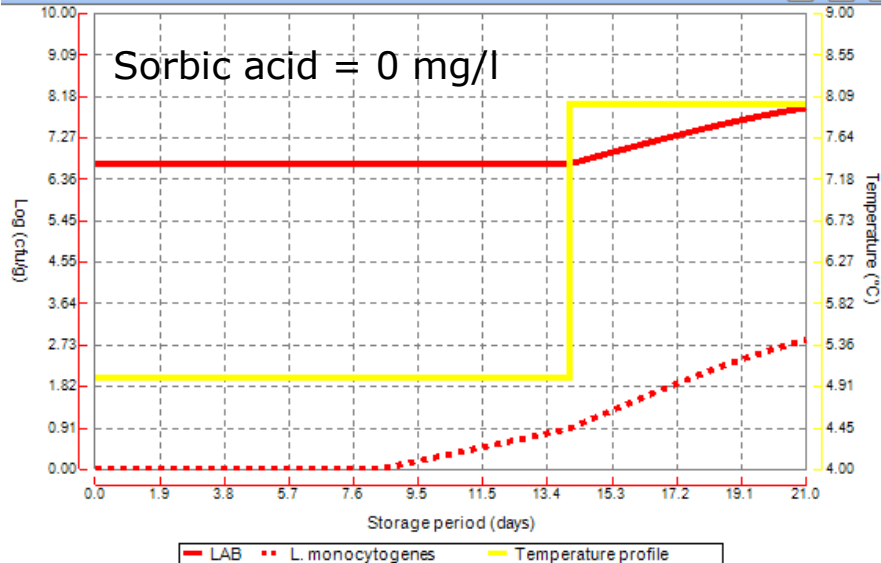
	Product 1	Product 2	Product 1	
<i>Listeria monocytogenes</i> , cfu/g	1	1	Temperature (°C)	7.0
Lactic acid bacteria, cfu/g	5000000	5000000	Salt in water phase of product, %	1.0
LAB aroma culture used	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	Initial pH	5.47
Worst case relative lag time	<input type="checkbox"/>	<input type="checkbox"/>	Lactic acid in water phase of product, mg/l	450
Storage period (d)	25		Sorbic acid in waterphase of product, mg/l	750

Apply Clear

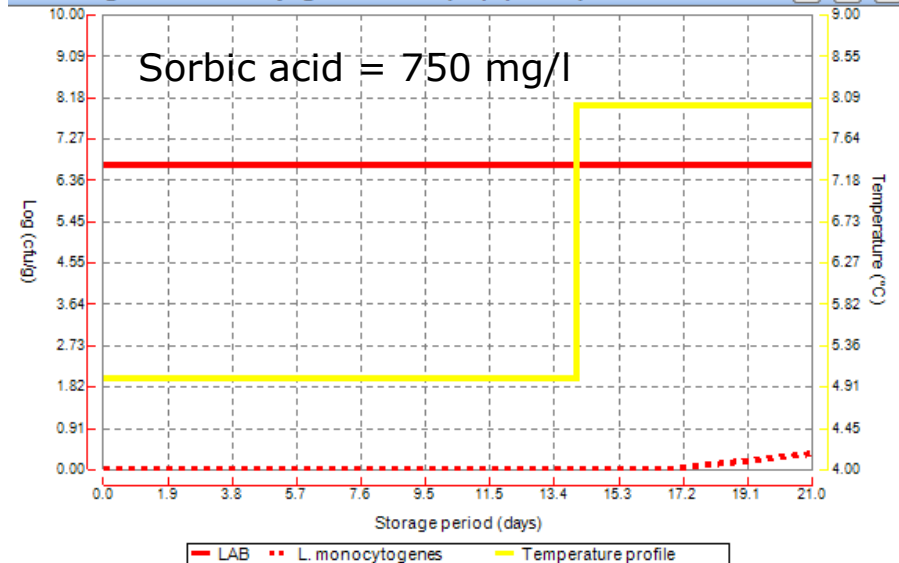
Constant temperature Series of constant temperatures Temperature profiles from logger data



Predicted growth of *L. monocytogenes* and LAB (temp. profile 1)

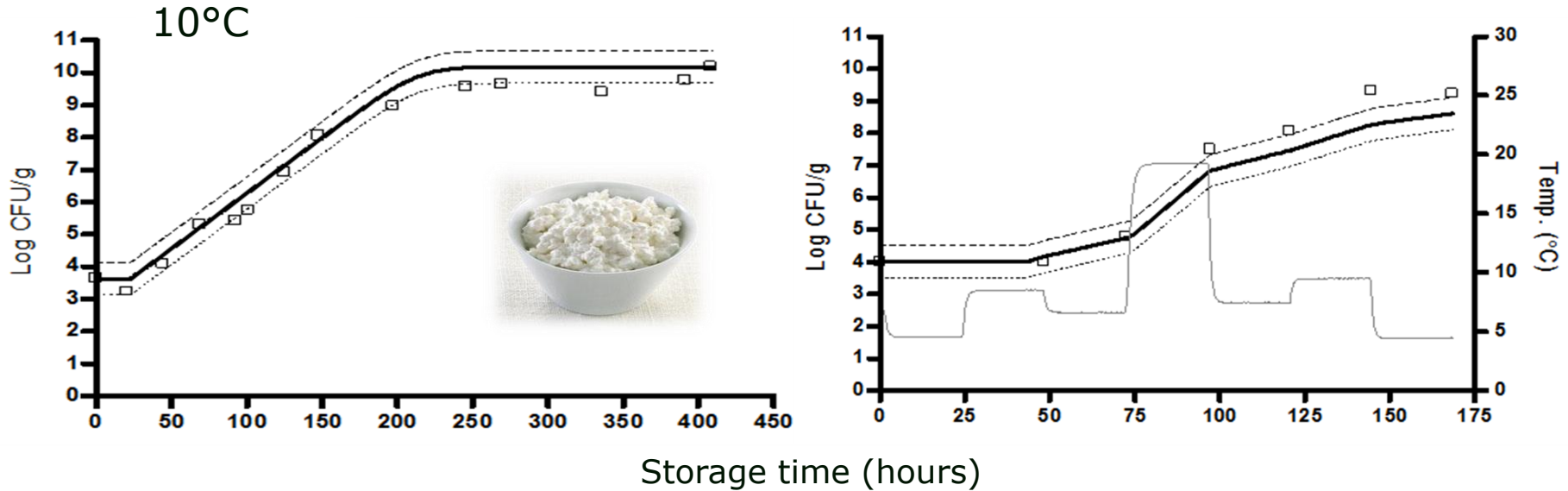


Predicted growth of *L. monocytogenes* and LAB (temp. profile 1)



Pseudomonas – Validation of growth model

Predictions are compared with growth measured in products at constant and varying storage temperatures



Predicted and observed growth of psychrotolerant *Pseudomonas* in cottage cheese with dressing with aroma culture

Generic growth model

Product characteristics

	Product 1	Product 2
Initial cell level (cfu/g)	1	1
Relative lag time (RTL)	3.1	3.1
Critical cells conc., log(cfu/g)	8.5	8.5
Max. pop. density, log(cfu/g)	9.5	9.5
Temperature (°C)	7.0	7.0
NaCl in water phase %	1.0	1.0
pH	5.30	5.47
Smoke components - phenol (ppm)	0	0
% CO2 in headspace gas at equilibrium	0	0
Nitrite, mg/kg	0	0

Organic acids in water phase of product

	Product 1	Product 2
Acetic acid (ppm)	0.0	0.0
Benzoic acid (ppm)	0	0
Citric acid (ppm)	0	0
Diacetate (ppm)	0	0
Lactic acid (ppm)	1700	450
Sorbic acid (ppm)	0	0
Storage period (d)	25	

Growth model selection

Select growth model

Pseudomonas - Cottage cheese - raw r

Delete model

Import model

Export model

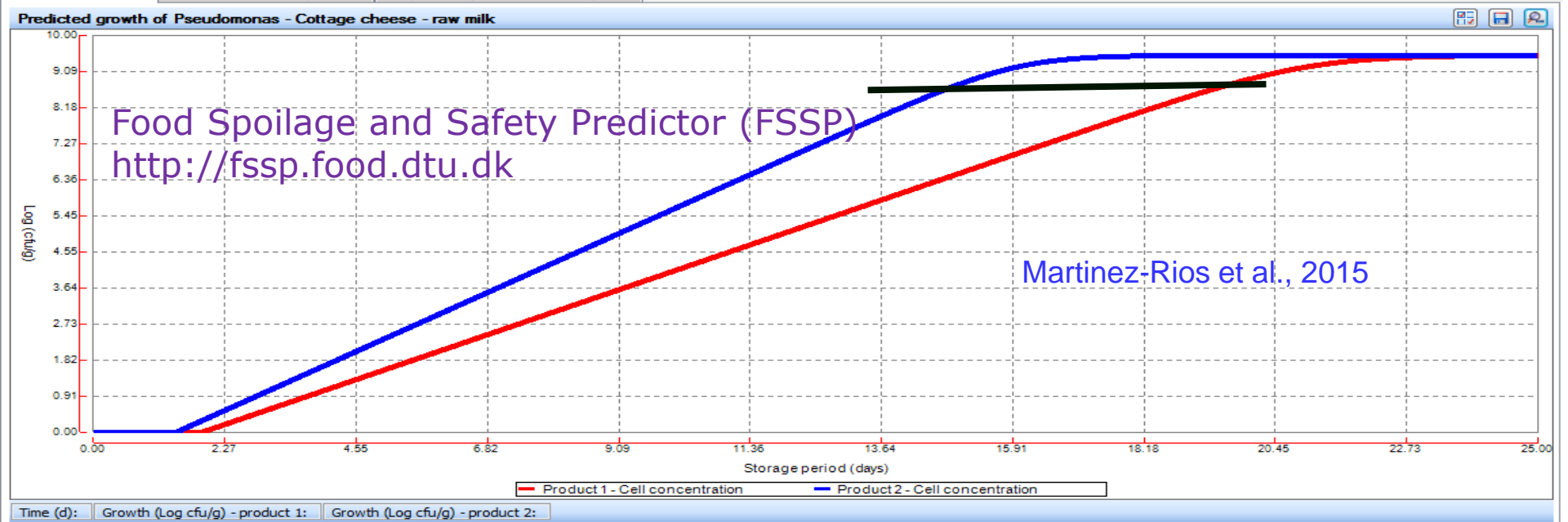
Edit model

Add model

Apply Clear



Constant temperature Series of constant temperatures Temperature, pH and lactic acid profiles



EU regulation (EC 2073/2005) and documentation



- Predictive mathematical modelling can be used to support documentation of compliance with microbiological criteria when based on physico-chemical product characteristics; storage and processing conditions, contamination and foreseen shelf-life
- Models are increasingly used by the seafood and meat sectors
- Development and validation of predictive models with a wide range of applicability have been more successful for the seafood and meat sectors than for the dairy sector

Models and software for *L. monocytogenes*

Type of model	Type of products	Factors	References
Growth	Liquid dairy products	T, pH, a_w , nitrite, CO ₂	Augustin et al. 2005
	Cheese	T, pH, a_w , nitrite, CO ₂	Augustin et al. 2005
	Ready-to-eat meat and seafood	T, pH, NaCl, phenol, nitrite, CO ₂ , AAC, DiAC, LAC	Mejlholm and Dalgaard, 2009
	Cottage cheese	T, pH, NaCl, LAC, SAC and LAB*	Østergaard et al. 2014
	Smear soft cheese - (past. or un-past. milk)	T, pH, a_w	Schvartzman et al. 2011
	Food	T, pH, a_w , LAC	Sym ´ Previous
	Cheese and dairy products	T, pH, a_w	te Giffel and Zwietering, 1999
Inactivation	Cheese	T, pH, a_w , LAC, SAC	Coroller et al. 2012

* Inhibiting effect of lactic acid bacteria (LAB)

Evaluation of models and software

Evaluation of *L. monocytogenes* models and software by using data collected from the scientific literature for different groups of cheeses



Prevalence of *Listeria monocytogenes* in European cheeses: A systematic review and meta-analysis

Veronica Martinez-Rios*, Paw Dalgaard

National Food Institute (DTU Food), Technical University of Denmark, Kgs. Lyngby, Denmark



Different groups of cheeses

References	Survey year	Number of <i>L. monocytogenes</i> positive (s) / total number of cheese samples (n)				
		Fresh	Ripened	Veined	Smear	Brined
Filiouis et al., 2009	2005-2006		4/20			0/10
Little et al., 2009	2006-2007		2/1240			0/10
O'Brien et al., 2009	2007	0/29	1/104		14/79	
Di Pinto et al., 2010	2007-2009					
Pesavento et al., 2010	2008					
Prencipe et al., 2010	2005-2006	1/437	1/449	21/444	24/802	
Angelidis et al., 2012	2010	0/83		0/38	0/16	
Lambertz et al., 2012	2006-2012				0/62	
Dambrosio et al., 2013	2009	0/404				
Doménech et al., 2013	2005-2009	0/77				
Parisi et al., 2013	2008-2010	3/70				
Gyurova et al., 2014	2011-2012			0/7		0/34
Doménech et al., 2015	2011-2012	2/100	5/100			
Schoder et al., 2015	NS ^a		1/15	0/50	1/22	
Spanu et al., 2015	2011-2013		3/50			7/33
Iannace et al., 2016	NS ^a	2/1	0/106	8/190	11/177	
Coronetti et al., 2016	NS ^a					15/87
Total		17/2580	15/2101	32/1218	50/1158	24/164



Brined (11.8% CI: 3.5-33.3)



Smear (5.1% CI: 1.9-13.1)



Veined (2.4% CI: 0.9-6.3)



Ripened (2.0% CI: 0.8-4.9)



Fresh (0.8% CI: 0.3-1.9)

Evaluation of growth models and software with literature data

Models are evaluated/validated by comparison of measured/observed and predicted values (kinetic parameters or shelf-life). The comparison can be graphical or mathematical

$$\text{Bias factor (B}_f\text{)} = 10^{\frac{\sum \log(\mu\text{-predicted} / \mu\text{-observed})}{n}}$$

$$\text{Accuracy factor (A}_f\text{)} = 10^{\frac{\sum |\log(\mu\text{-predicted} / \mu\text{-observed})|}{n}}$$

Conclusions

- Models for growth of *L. monocytogenes* in dairy products often include the effect of temperature, pH, salt/aw and some organic acids
 - **Østergaard et al., 2014** included the inhibiting effect of LAB and we **recommend** to use this model when **LAB are present**
- Cardinal parameter model including the effect of specific dairy components (**melting salts and gluconic acid**) have been developed for *Listeria monocytogenes* in an on going project DAIRY-PREDICT (2015-2019)
 - the model can be used for **re-formulation** of product, simulation of **storage conditions** and **documentation of safety**

- Anti-listerial compounds in fermented dairy products (**peptides and bacteriocins**) are interesting to include in extended model
- To benefit from the potential of predictive models further developments are needed within the dairy sector
 - Models for human pathogens other than *L. monocytogenes*
 - Collaboration between processors, culture-producers and scientists to include effects of dairy specific factors in validated models
- Help to establish safe shelf-life