

# ***Food Safety and Quality under the Auspices of Data Science***

## ***IAFP : Webinar 24/10/2023***

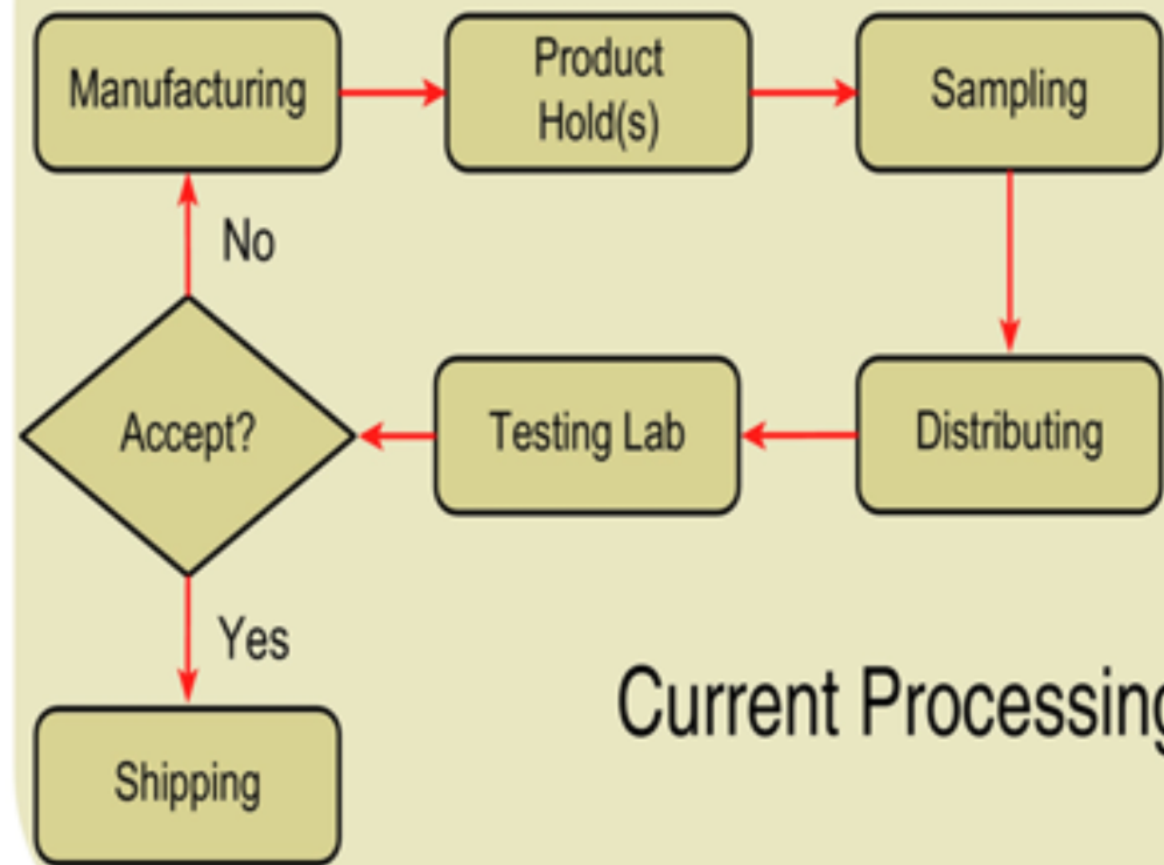


**George-John NYCHAS**  
**Agricultural University of Athens -**  
**Greece**



# Predictive Modeling





Current Processing

## I

*(Acts whose publication is obligatory)*

**COMMISSION REGULATION (EC) No 2073/2005**  
**of 15 November 2005**  
**on microbiological criteria for foodstuffs**  
**(Text with EEA relevance)**

THE COMMISSION OF THE EUROPEAN COMMUNITIES,

Having regard to the Treaty establishing the European Community,

- (4) Microbiological criteria also give guidance on the acceptability of foodstuffs and their manufacturing, handling and distribution processes. The use of microbiological criteria should form an integral part of the implementation of HACCP-based procedures and other hygiene control measures.

# (EC) No 1441/2007 of 5 December 2007 amending Regulation (EC) No 2073/2005 on microbiological criteria for foodstuffs

## Chapter 2. Process hygiene criteria 2.1 Meat and products thereof

Carcases of cattle, sheep, goats and <b>horses</b>	Aerobic colony count 3,5 to 5,0 log cfu/cm <sup>2</sup>	Carcases after dressing but before chilling
	Enterobacteriaceae 1,5 to 2,5 log cfu/cm <sup>2</sup>	
Minced beef 5 × 10 <sup>5</sup> - 5 × 10 <sup>6</sup> cfu/g	Aerobic colony count End of the manufacturing process	

## LEGISLATION



Minced beef  
5 × 10<sup>5</sup> - 5 × 10<sup>6</sup> cfu/g

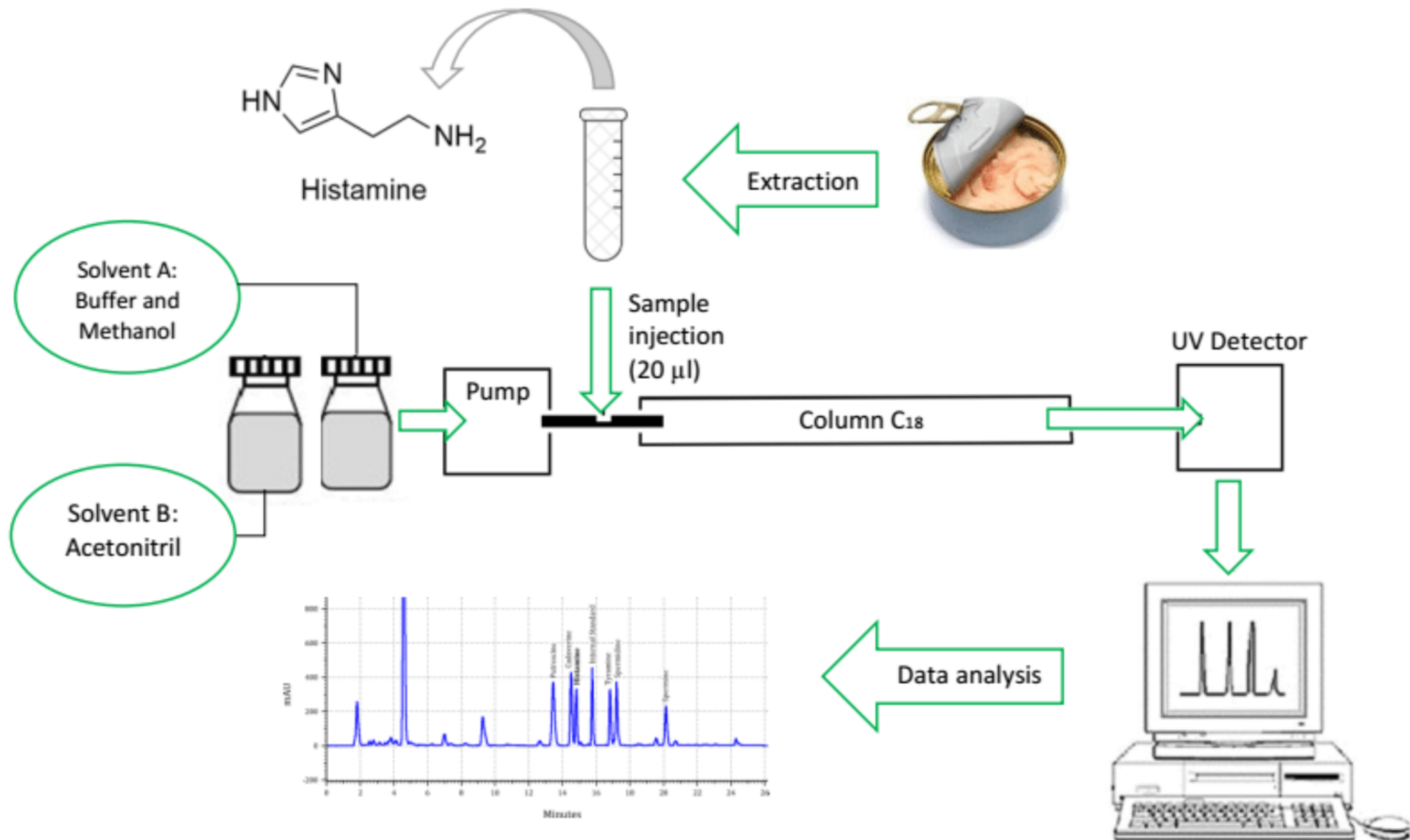
Aerobic colony count  
End of the manufacturing process

LEGISLATION

28/02/2022

6

Food category	Micro-organisms/their toxins, metabolites	Sampling-plan <sup>(1)</sup>		Limits <sup>(2)</sup>		Analytical reference method <sup>(3)</sup>	Stage where the criterion applies
		n	c	m	M		
1.26. Fishery products which have undergone enzyme maturation treatment in brine, manufactured from fish species associated with a high amount of histidine <sup>(16)</sup>	Histamine	9	2	200 mg/kg	400 mg/kg	HPLC <sup>(18)</sup>	Products placed on the market during their shelf-life





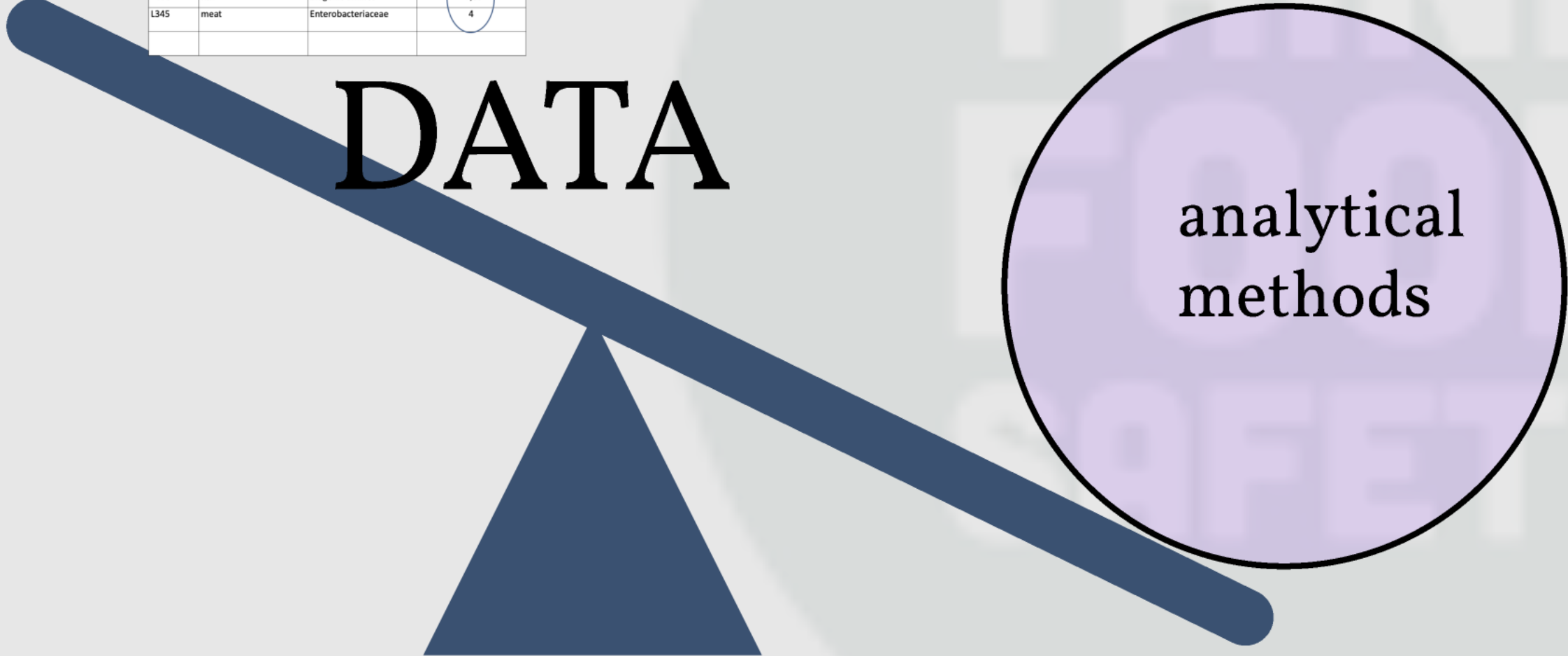
**The analysis of THE END / FINISHED product does provide a SINGLE number on which the (whole) production process will be assessed**

<b>sample code</b>	<b>Sample description (e.g. pork, chicken, dairy etc.)</b>	<b>Hazards / Quality index</b>	<b>population/concentration (eg. cfu/g, mg/Kg)</b>
K098	cream	<i>Listeria monocytogenes</i>	55
B079	fish	biogenic amines	17,35
L345	meat	Enterobacteriaceae	4

The analysis of THE END / FINISHED product does provide a SINGLE number on which the (whole) production process will be assessed

sample code	Sample description (e.g. pork, chicken, dairy etc.)	Hazards / Quality Index	population/concentration (eg. cfu/g, mg/kg)
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# DATA



analytical  
methods

## **Article 3 & 2; 2073/2005:**

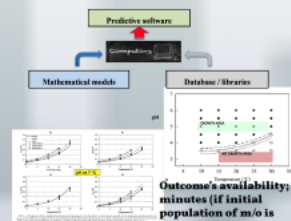
As necessary, the food business operators responsible for the manufacture of the product shall conduct studies in accordance with **Annex II** in order to investigate compliance with the criteria throughout the shelf-life.

# The role of Predictive Modeling in Food Safety & Quality

**What is Predictive Modeling?**  
 Predictive modeling is a statistical technique that uses historical data to predict future outcomes. In the context of food safety and quality, it can be used to predict the growth of pathogens, the shelf life of products, and the impact of various factors on food quality.

**Why is Predictive Modeling Important?**  
 Predictive modeling is important because it allows food safety and quality professionals to make data-driven decisions. By understanding the factors that influence food safety and quality, they can develop strategies to prevent problems before they occur.

**How is Predictive Modeling Used?**  
 Predictive modeling is used in a variety of ways in the food industry. For example, it can be used to predict the growth of pathogens in different environments, to estimate the shelf life of products, and to identify the factors that contribute to food quality issues.



**TABLE 1. Specific growth rates of *L.* and *S.* growing in minimal medium at different temperatures and storage times as determined previously by DSC 4 model and Baranyi and Ratkowsky models.**

Source: Advanced Microbial Pathology Laboratory, Chen, Hengshen, J. Weiss, R. Ratkowsky.

Baranyi and Ratkowsky model

Strain	Storage time	Specific growth rate (1/h)	Log phase growth rate (1/h)
Clostridia	5°C	0.078	0.078
	10°C	0.477	0.477
Listeria	5°C	0.048	0.048
	10°C	0.477	0.477



**USST (China) Microrisk Lab**  
 an online modeling framework for predictive microbiology

**Available 47**

**4 Chemical and Physical Database**

**4 Integrated Predictive Models**

**4 Statistical Evaluation**

**Increasing of Log in Karthi (cells/g) change at different To C**  
 (Temperature of Storage of Food at 10°C and 15°C)

Time (h)	10°C	15°C	20°C	25°C	30°C
0	0.00	0.00	0.00	0.00	0.00
1	0.01	0.02	0.03	0.04	0.05
2	0.02	0.04	0.06	0.08	0.10
3	0.03	0.06	0.09	0.12	0.15
4	0.04	0.08	0.12	0.16	0.20
5	0.05	0.10	0.15	0.20	0.25
6	0.06	0.12	0.18	0.24	0.30
7	0.07	0.14	0.21	0.28	0.35
8	0.08	0.16	0.24	0.32	0.40
9	0.09	0.18	0.27	0.36	0.45
10	0.10	0.20	0.30	0.40	0.50

**Contribution of individual strains in the cocktail experiments at the end of storage of karthi change stored at different temperatures**

**% contribution of *Listeria monocytogenes* strains**

Temperature °C	TS124	TS125	TS128	TS131	TS133
5°C	0	0	0	100	0
10°C	0	95	0	0	5
15°C	18	68	0	14	0
20°C	0	78	5	3	14

Failure (%) of models to describe the data at 50°C (but NN)

**Food Microbiology**

Strain	Temperature (°C)	Time (h)	Log CFU/g
Listeria	5	0	0.00
	10	0	0.00
	15	0	0.00
	20	0	0.00
Salmonella	5	0	0.00
	10	0	0.00
	15	0	0.00
	20	0	0.00



**Food Safety and Quality**

**Food Safety and Quality**

**Food Safety and Quality**

# The role of Predictive Modeling in Food Safety & Quality

## ANNEX II

The studies referred to in Article 3(2) shall include:

- specifications for physico-chemical characteristics of the product, such as pH, aw, salt content, concentration of preservatives and the type of packaging system, taking into account the storage and processing conditions, the possibilities for contamination and the foreseen shelf-life, and
- consultation of available scientific literature and research data regarding the growth and survival characteristics of the micro-organisms of concern.

When necessary on the basis of the above mentioned studies, the food business operator shall conduct additional studies, which may include:

- predictive mathematical modeling established for the food in question, using critical growth or survival factors for the micro-organisms of concern in the product,
- tests to investigate the ability of the appropriately inoculated micro-organism of concern to grow or survive in the product under different reasonably foreseeable storage conditions,
- studies to evaluate the growth or survival of the micro-organisms of concern that may be present in the product during the shelf-life under reasonably foreseeable conditions of distribution, storage and use.

The above mentioned studies shall take into account the inherent variability linked to the product, the microorganisms in question and the processing and storage conditions.

Predictive software

Computing

Mathematical models

Database / libraries

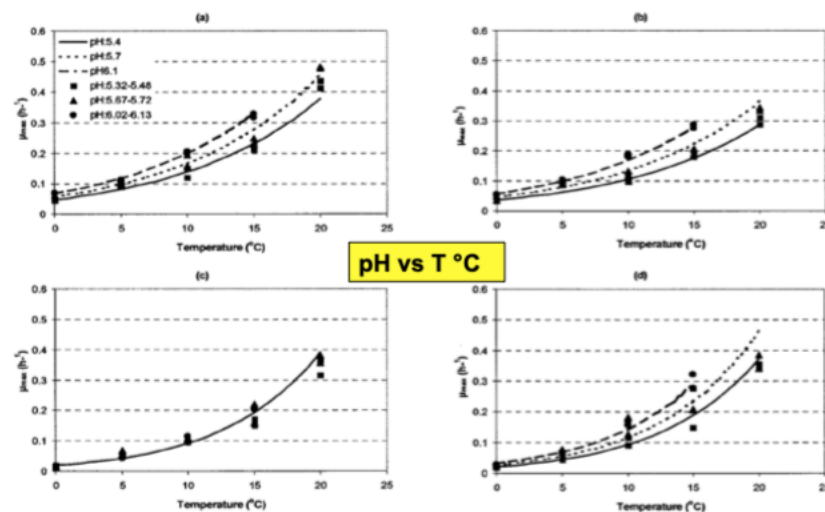
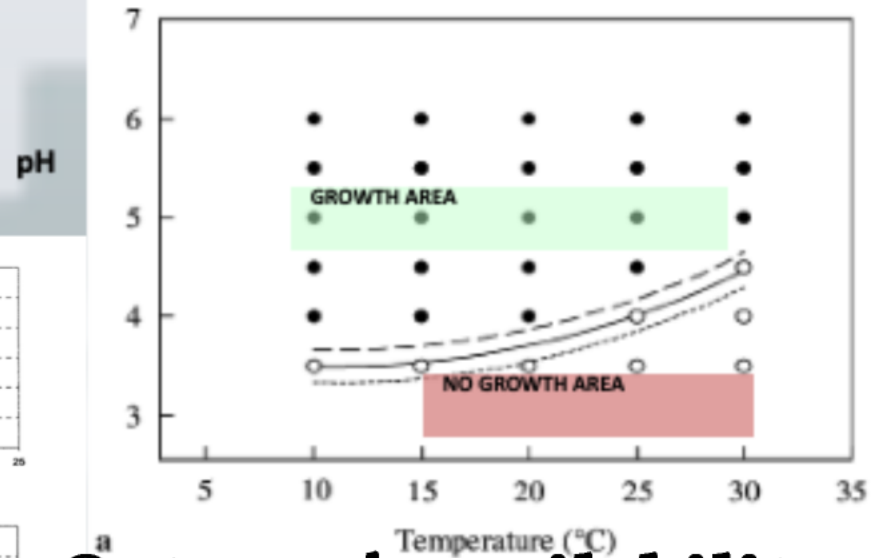
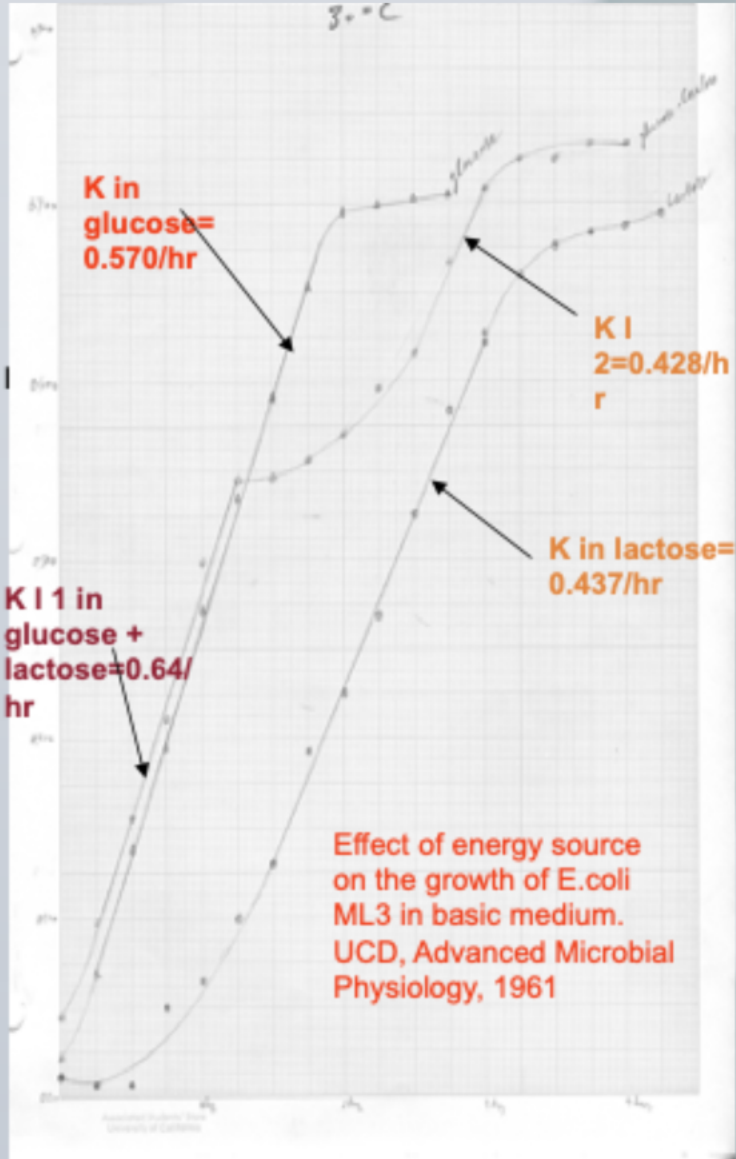


FIG. 4. Predictions of the modified Arrhenius model (equation 1) for the effect of temperature and pH on the maximum specific growth rate ( $p_{max}$ ) of the different spoilage bacteria (a, pseudomonads; b, *Brochothrix thermophilus*; c, lactic acid bacteria; d, *Enterobacteriaceae*) on ground meat. Lines represent predictions of equation 1 at three different initial pH values of meat. Points represent observed values of  $p_{max}$ .



**Outcome's availability;  
minutes (if initial  
population of m/o is  
known)**





**TABLE 1. Specific growth rates of *E. coli* ML3 growing in minimal medium with three different energy sources as determined graphically by TINO'S hand and Baranyi and Roberts (1994).**

**Source:** Advanced Microbial Physiology Laboratory Class. University of California, Davis 1961 (Professors: J. Ingraham, J. Marr, R. Hungate)

Energy source	Determination Method		
	Graphical 1961	Baranyi and Roberts 1994	
	Specific growth rate (1/h)	Specific growth rate (1/h)	Lag phase (h)
Glucose	0.570	<b>0.619</b>	0.088
Lactose	0.437	<b>0.461</b>	0.905
Glucose + Lactose Phase 1	0.640	<b>0.580</b>	0.000
Glucose + Lactose Phase 2	0.428	<b>0.368</b>	0.763





Contents lists available at [ScienceDirect](#)

## Food Microbiology

journal homepage: [www.elsevier.com/locate/fm](http://www.elsevier.com/locate/fm)



### Software for predictive microbiology and risk assessment: A description and comparison of tools presented at the ICPMF8 Software Fair

Fanny Tenenhaus-Aziza <sup>a</sup>, Mariem Ellouze <sup>b, \*</sup>

Software	Accessibility	Date of creation	Targeted users	Modeling approach
Baseline	Free, internet access	2012	FBOs, Researchers, Teachers, Students	Deterministic
ComBase	Free, internet access	2004	FBOs, Researchers, Teachers, Students, Government	Deterministic
Dairy Products Safety Predictor	Commercial, internet access	2012	FBOs	Probabilistic
FDA-iRISK	Free, internet access	2012	FBOs, Researchers, Teachers, Students, Government	Probabilistic
FILTREX	Free, downloadable	2013	Researchers, Teachers, Students	Probabilistic
FISHMAP	Free, downloadable	2011	FBOs, Researchers, Teachers, Students, Government	Deterministic
Food Spoilage and Safety Predictor (FSSP)	Free, downloadable	1999	FBOs, Researchers, Teachers, Students, Government	Deterministic
GInaFiT	Free, downloadable	2003	FBOs, Researchers, Teachers, Students, Government	Deterministic
GroPIN	Free, downloadable	2013	FBOs, Researchers, Teachers, Students, Government	Deterministic and probabilistic
Listeria Meat Model	Commercial, downloadable	2012	FBOs, Government	Deterministic
MicroHibro	Free internet access	2011	FBOs, Researchers, Teachers, Students, Government	Probabilistic
MRV, Microbial Responses Viewer	Free internet access	2008	FBOs, Researchers, Teachers, Students, Government	Deterministic
NIZO Premia	Commercial, no internet access	1995	FBOs	Deterministic
PMM-Lab	Free, internet access	2012	Researchers, Teachers, Students, Government	Deterministic
Prediction of Microbial Safety in Meat Products	Free, internet access	2006	FBOs, Researchers, Teachers, Students, Government	Deterministic
Sym'Previous	Commercial, internet access	2003	FBOs, Researchers, Teachers, Students, Government	Deterministic and Probabilistic

# USST (China) Microrisk Lab

*an online modeling freeware  
for predictive microbiology*

AVAILABLE AT



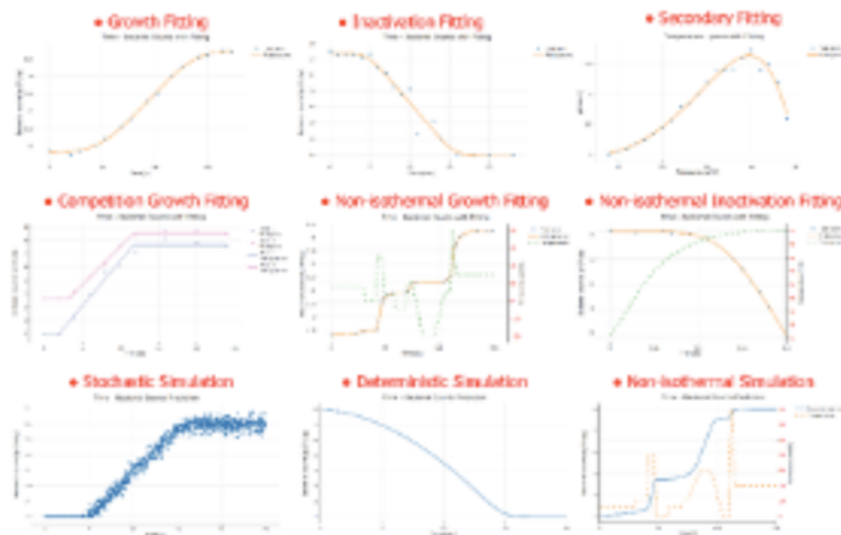
<https://www.microrisklab.top/chinese/>  
<https://www.microrisklab.top/english/>



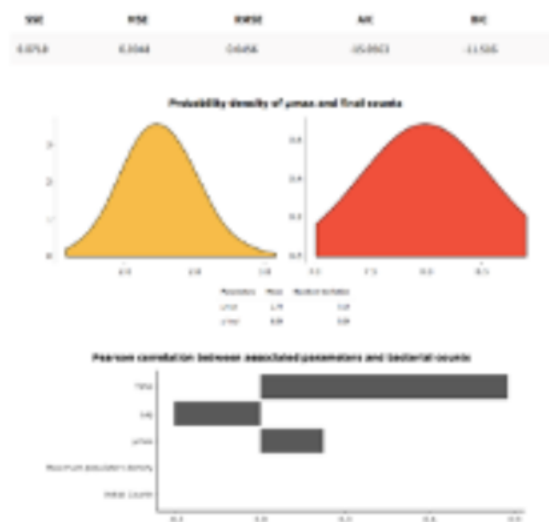
## ★ Numerical and Graphical Interface



## ★ Integrated Predictive Models



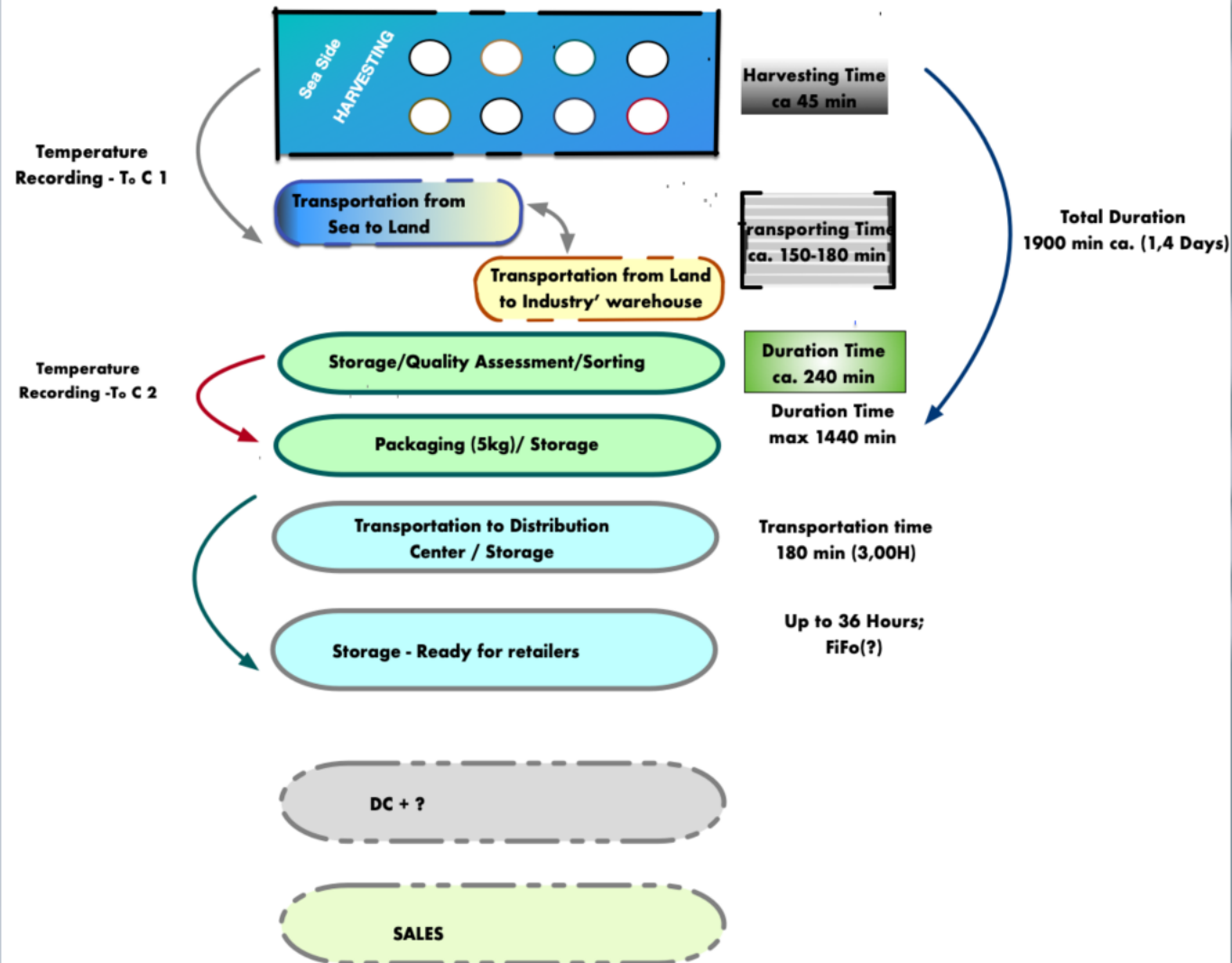
## ★ Statistical Indicators







## Monitoring T C, from Primary Production (fish) to retailers





# SQP Expiration Date Calculator

The Lot number field is required to calculate results  
The Bin EPC input field is optional

Lot Number

029099

Bin EPC

CFU

3

Sample Hou...

1

Count

200

Generator Type

Indicative

Reference L...

999999

Calculate

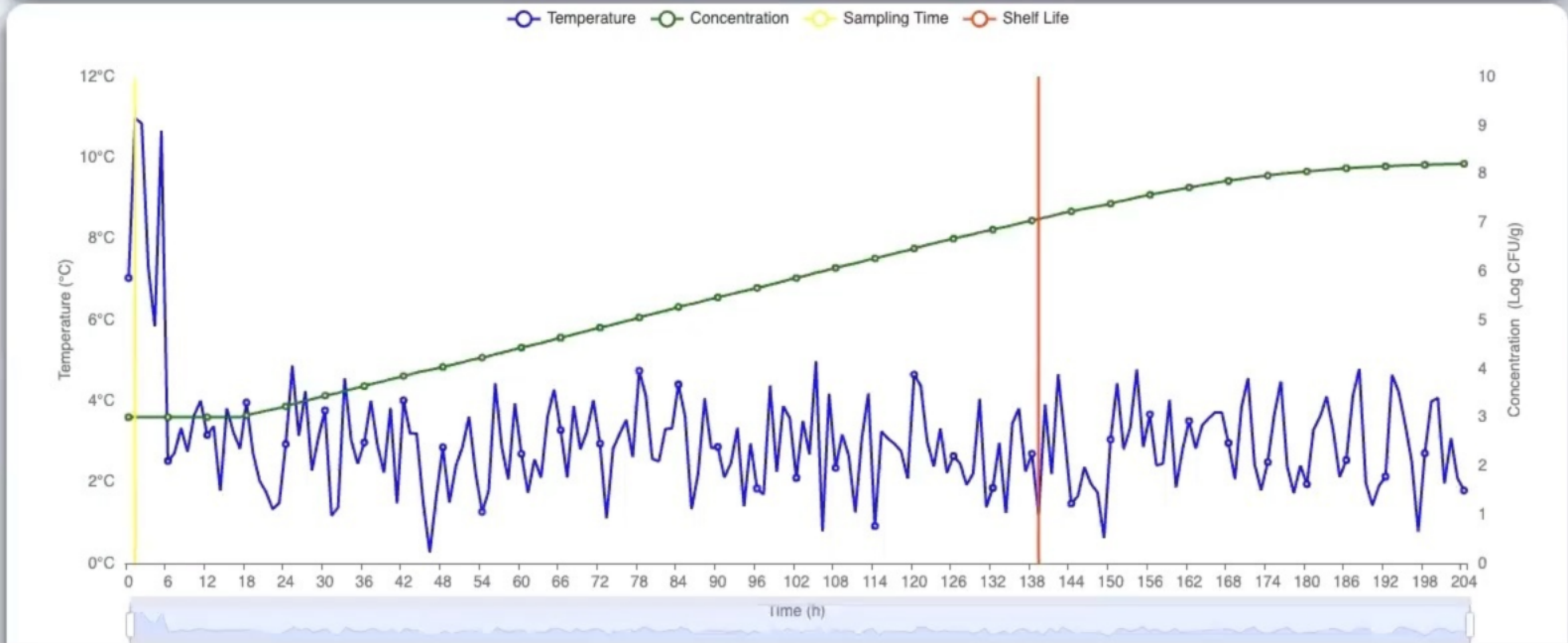
Shelf Date:: 15/11/2022

Start Date:: 09/11/2022

End Date:: 17/11/2022

Shelf Life:: 139

Sampling Hour:: 1



Calculation Results		
#	T °C	Population
0	7.03°C	3.0
1	10.97°C	3.0
2	10.84°C	3.0
3	7.31°C	3.0
4	5.84°C	3.0
5	10.66°C	3.0
6	2.52°C	3.0
7	2.71°C	3.0
8	3.33°C	3.0
9	2.74°C	3.0
10	3.64°C	3.0
11	4.01°C	3.0

# Survival of Lm in Katiki (soft) cheese at different To C

Panagou 2008 JFP; Mataragkas et al 2008 JFP; Kagkli et al AEM 2009

TABLE 1. Parameter estimation and statistical indices of the different models used for fitting the experimental data<sup>a</sup>

Model type	$N(0)$ (log CFU g <sup>-1</sup> )	$A^b$ (log CFU g <sup>-1</sup> )	$N_{res}$ (log CFU g <sup>-1</sup> )	$k_{max}$ (day <sup>-1</sup> )	$t_s$ (days)	$\delta$ (days)	$p$	RMSE	$R^2$	$f$
<b>rGompertz<sup>c</sup></b>										
5°C	6.13 ± 0.24	4.21 ± 0.31		0.71 ± 0.14	4.96 ± 1.55			0.371	0.967	4.62 <sup>d</sup>
10°C	5.98 ± 0.24	4.35 ± 0.32		0.62 ± 0.11	3.87 ± 0.96			0.334	0.985	1.35
15°C	5.96 ± 0.12	4.34 ± 0.18		0.87 ± 0.09	3.46 ± 0.34			0.249	0.992	0.93
20°C	5.71 ± 0.21	4.38 ± 0.32		1.37 ± 0.32	2.85 ± 0.55			0.422	0.978	2.13
<b>mWeibull<sup>e</sup></b>										
5°C	6.26 ± 0.25		1.96 ± 0.13			5.81 ± 1.22	1.88 ± 0.67	0.336	0.976	3.78 <sup>d</sup>
10°C	6.22 ± 0.23		1.71 ± 0.13			4.38 ± 0.84	1.44 ± 0.27	0.265	0.985	0.85
15°C	6.13 ± 0.13		1.71 ± 0.11			4.10 ± 0.37	1.93 ± 0.25	0.186	0.992	0.52
20°C	6.01 ± 0.25		1.45 ± 0.17			2.87 ± 0.47	1.94 ± 0.43	0.236	0.989	0.67
<b>Geeraerd</b>										
5°C	6.26 ± 0.23		1.88 ± 0.13	0.81 ± 0.14	2.84 ± 1.41			0.301	0.981	3.05 <sup>d</sup>
10°C	6.11 ± 0.21		1.69 ± 0.13	1.01 ± 0.11	2.70 ± 0.92			0.245	0.987	0.73
15°C	6.02 ± 0.09		1.71 ± 0.08	1.61 ± 0.12	3.01 ± 0.30			0.145	0.995	0.32
20°C	5.87 ± 0.25		1.43 ± 0.20	2.38 ± 0.40	2.22 ± 0.54			0.357	0.976	1.52
<b>RBF NN<sup>f</sup> Radial Basic Function Neural Network (RBF NN)</b>										
5°C								0.218	0.992	1.59
10°C								0.216	0.993	0.57
15°C								0.155	0.996	0.36
20°C								0.265	0.989	0.84

Failure (?) of models to describe the data at 50 C (but NN)

## Contribution of individual strains in the cocktail experiments at the end of storage of katiki cheese stored at different temperatures

Temperature °C	% contribution of <i>Listeria monocytogenes</i> strains				
	TS124	TS125	TS128	TS131	TS133
5°C	0	0	0	100	0
10°C	0	95	0	0	5
15°C	18	68	0	14	0
20°C	0	78	5	3	14

Provisional text

JUDGMENT OF THE COURT (Eighth Chamber)

30 June 2022 (\*)

(Reference for a preliminary ruling – Food law– Regulation (EC) No 2073/2005 – Microbiological criteria for foodstuffs – Article 3(1) – Obligations of food business operators – Annex I – Point 1.2 of Chapter 1 – Limit values for the presence of *Listeria monocytogenes* in fish products before and after being placed on the market – Regulation (EC) No 178/2002 – Article 14(8) – Official controls of the product at the stage at which it is placed on the market – Scope)

In Case C-51/21,

REQUEST for a preliminary ruling under Article 267 TFEU from the Tallinna Halduskohus (Administrative Court, Tallinn, Estonia), made by decision of 28 January 2021, received at the Court on 28 January 2021, in the proceedings

On those grounds, the Court (Eighth Chamber) hereby rules:

**The combined provisions of Article 3(1) and point 1.2 of Chapter 1 of Annex I to Commission Regulation (EC) No 2073/2005 of 15 November 2005 on microbiological criteria for foodstuffs, as amended by Commission Regulation (EU) 2019/229 of 7 February 2019, must be interpreted as meaning that, where the manufacturer is unable to demonstrate, to the satisfaction of the competent authority, that, throughout their shelf-life, foodstuffs will not exceed the limit of 100 colony-forming units/grams (g), as regards the presence of *Listeria monocytogenes*, the limit requiring the absence of detection of *Listeria monocytogenes* in 25 g of the food product concerned laid down in that point 1.2 of that Annex I, *does not apply to foodstuffs which have been placed on the market throughout their shelf-life.***

[Signatures]

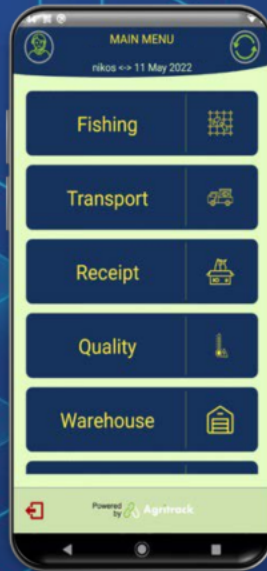


# FISH-TRACK

## ELEVATING AQUACULTURES

### TRACEABILITY AUTOMATION RECORD & SECURE

01



IOT & LEDGER

Digitize, track & improve operations

### QUANTIFIABLE QUALITY MONITORING

02

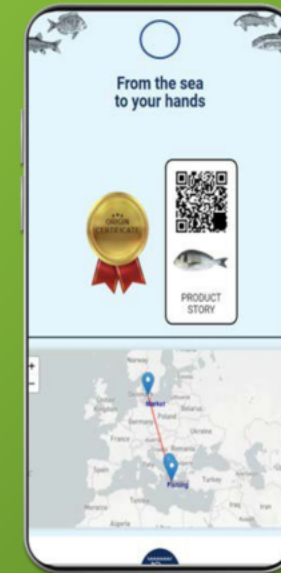


Predictive Microbiology  
Artificial Intelligence & NIR

Assess fish expiration date, safety and freshness

### FOOD VALUE CHAIN INTELLIGENCE INFORM CONSUMERS

03



QR Code

Enhance marketing  
Show provenance and quality



[Logout](#) [About](#)

Lot: 029099

Fish Date: 02/06/2023

Timezone: UTC+03:00

Species: BREAM

Batch #: 1705231

# Bins: 10

Min Temp: 1.69°C

Max Temp: 10.97°C

Span: 168 hours

Estimated Shelf Life:

Action	Date	Location
Hatched	08/12/2020 03:34	FARM 1
Fished	02/06/2023 06:39	FARM 1
Transport	02/06/2023 07:45	FARM 1
Plant	02/06/2023 10:49	PLANT 1
QC	02/06/2023 11:12	PLANT 1
Packaged	02/06/2023 11:18	PLANT 1
Dispatched	02/06/2023 16:21	PLANT 1
Received	02/06/2023 20:25	DISTRIBUTION 1
Shipped	03/06/2023 13:22	DISTRIBUTION 1
Delivered	09/06/2023 09:55	RETAIL 1
Scanned	09/06/2023 22:09	RETAIL 1

Temps/Lot



Indicative shelf life graph



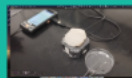


## Coming Trends

Having in mind paragraph (24) of EU 2073/2009

(24) ..... food business operators should have the possibility to use analytical methods other than the reference methods, in particular more rapid methods, as long as the use of these alternative methods provides equivalent results. Moreover, a sampling plan needs to be defined for each criterion in order to ensure harmonised implementation.

It is nevertheless necessary to allow the use of other sampling and testing schemes, including the use of alternative indicator organisms, on condition that these schemes provide equivalent guarantees of food safety.



Data Acquisition [food spoilage, drug resistance, food authenticity]



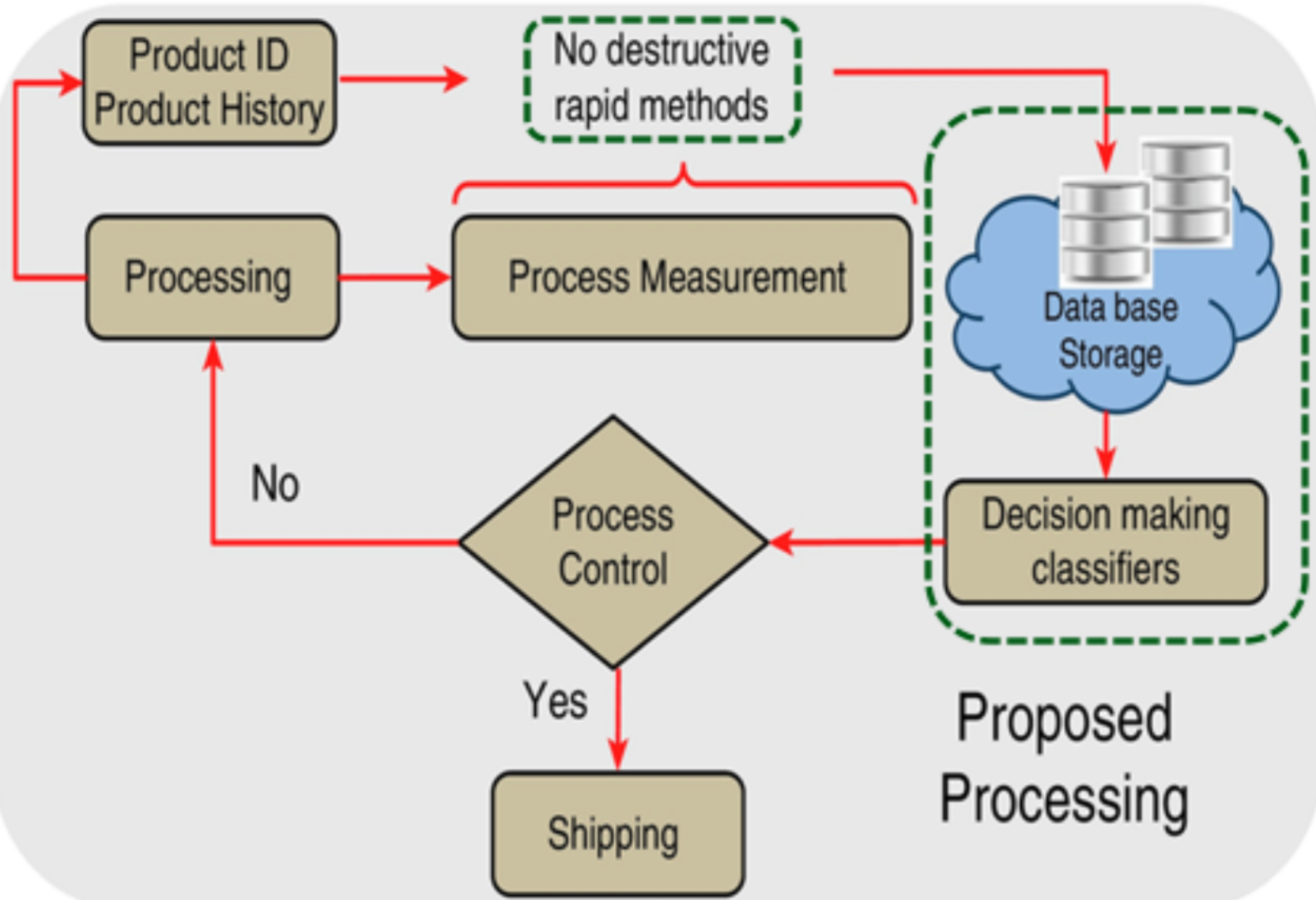
# Coming Trends

THINK  
FOOD  
SAFETY

Having in mind paragraph (24) of EU 2073/2005

(24) ..... food business operators should have the possibility to use analytical methods other than the reference methods, in particular more rapid methods, as long as the use of these alternative methods provides equivalent results. Moreover, a sampling plan needs to be defined for each criterion in order to ensure harmonised implementation.

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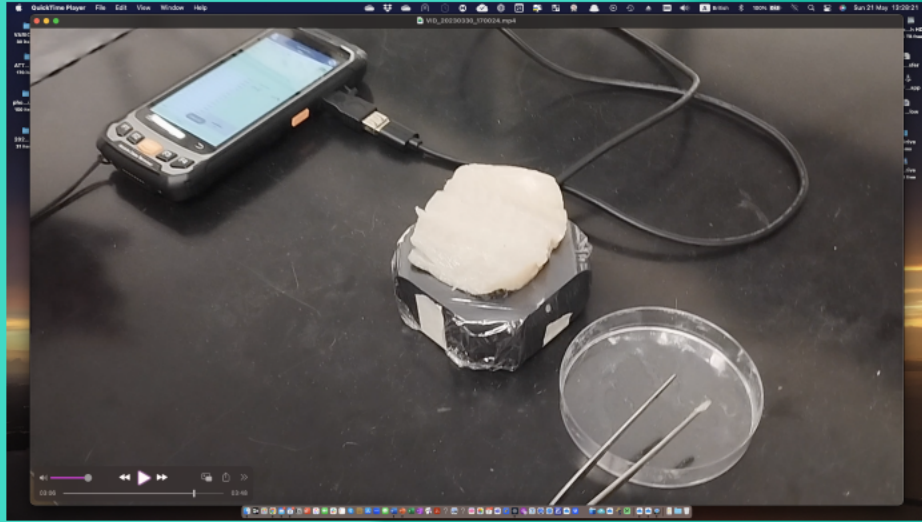






SAFETY





**In total there have been collected about ca 25000 measurements from 12000 samples [Vibrational Spectroscopy & Machine Learning analysis]**

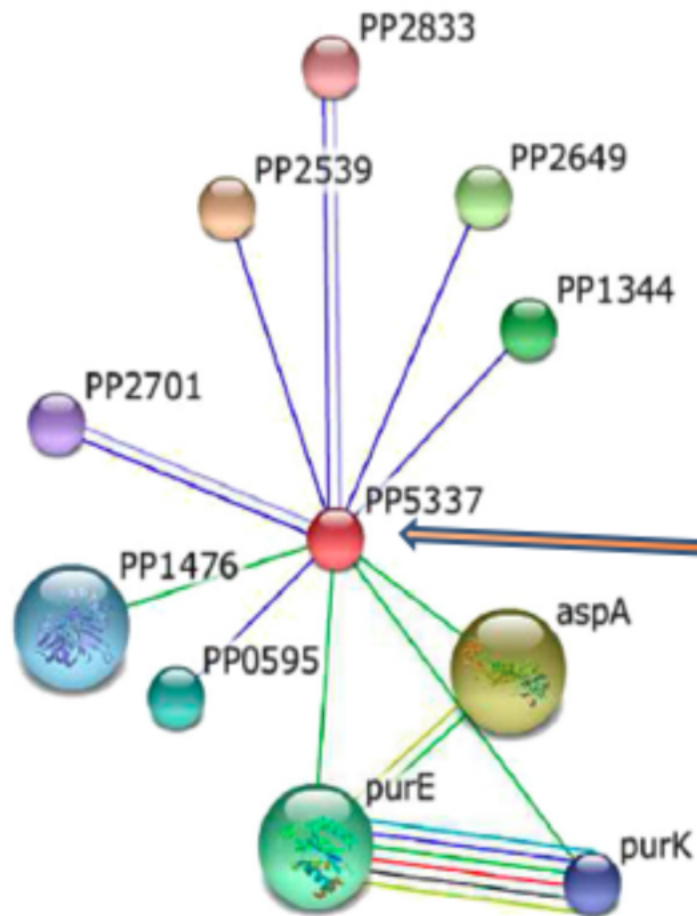
**Data Acquisition [food spoilage, drug residues, food pathogens]**



# Next Generation Sequencing



## Identification of meat spoilage gene biomarkers in *Pseudomonas putida* using gene profiling



Gene "PP5337" is found to be directly linked with "aspA" (aspartate ammonia-lyase) and "purE" (phosphoribosylaminoimidazole regulator), indicating that it can also be associated with the malodours production i.e aerobic spoilage

Gene "PP5337"

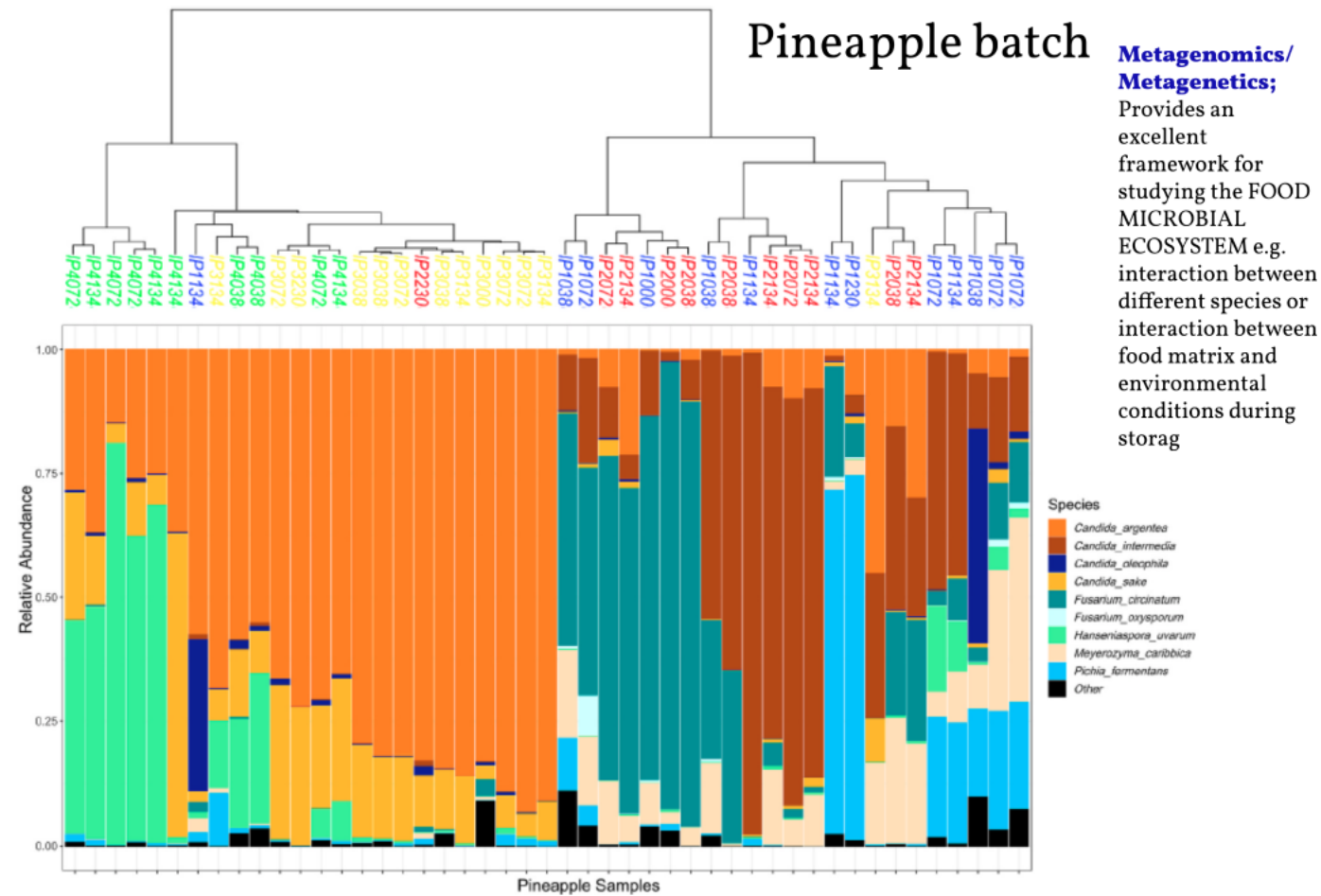
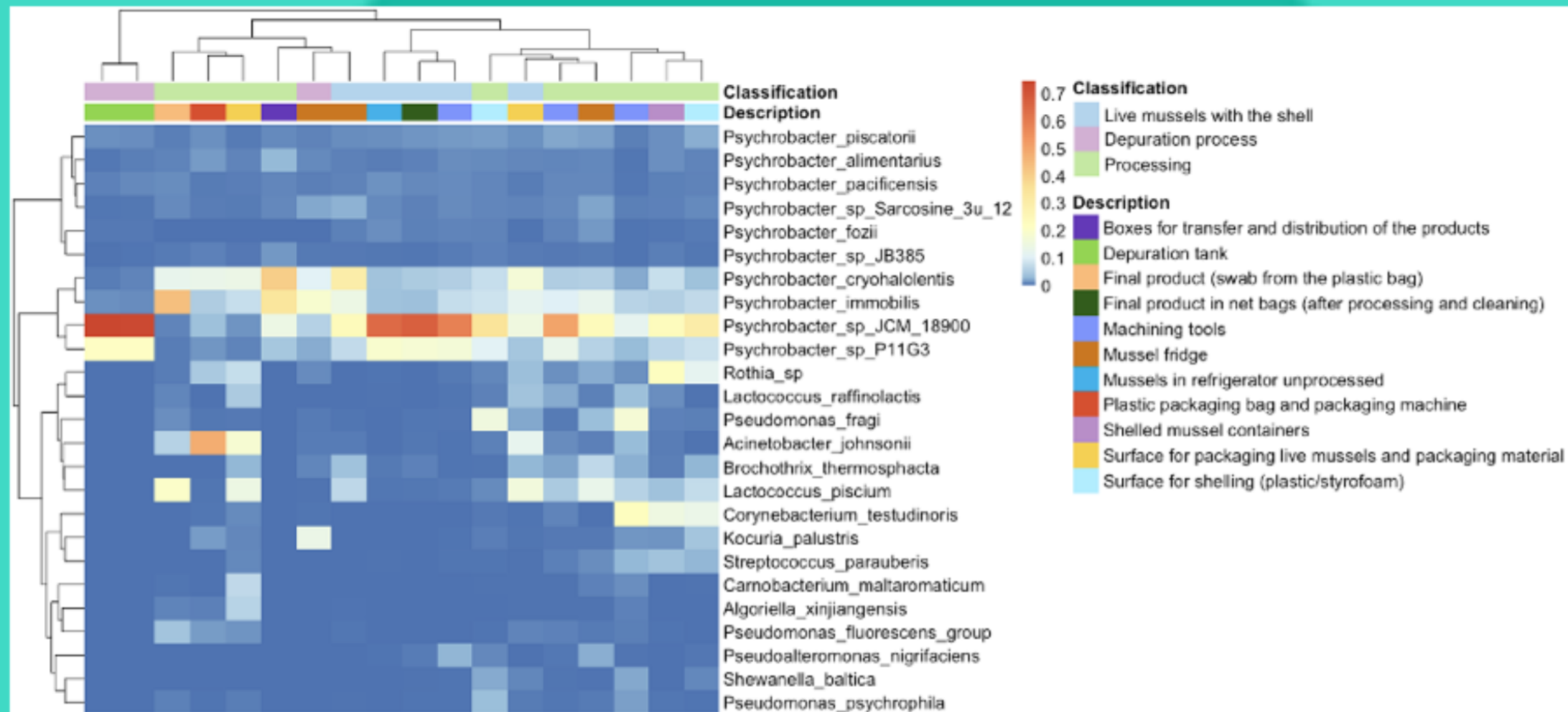


Fig. 5. Composition plot showing the relative abundances of the nine main Ascomycota species found in Pineapples samples. On the top: hierarchical clustering of batches samples according to Bray-Curtis distance and ward algorithm (blue for P1, red for P2, yellow for P3 and green for P4 as shown in Fig. 4). (For interpretation of the references to colour in this figure legend, the reader is referred to the Web version of this article.)

## Mussels processing plant – Batch 2 or 5 - UNINA






**WP5: Risk assessment of biological, chemicals hazards across food chain in EU and China (SCU)**



## Macrogenomic Analysis Platform: Flow path and Demo


Metagenomic data analysis

10.6.144.190:8000



# 宏基因组数据分析

## Metagenomic Data Analysis



I Begin


II Email

III Option

IV Upload

V Run

### Begin



This is a metagenomic data analysis platform, based on metaWRAP, which can help you with data quality control, species annotation and visualization, metagenomic assembly, three mainstream binning methods, bin reassembly, and annotation. You can upload your data, choose the analysis modules you want to perform, and then wait for the results to be completed.

ation. You can upload your  
modules you want to per-  
results to be completed.

## WP5: Risk assessment of biological, chemicals hazards across food chain in EU and China (SCU)



四川大学  
SICHUAN UNIVERSITY

### Macrogenomic Analysis Platform: Flow path and Demo

Metagenomic data analysis

10.6.144.190:8000/run



宏基因组数据分析  
Metagenomic Data Analysis



I Begin

II Email

III Option

IV Upload

V Run

Please confirm your task information

Your email: 2678190754@qq.com

Your data: test2\_1.fastq,test2\_2.fastq

Your option: kraken2,blobology,bin\_quantification,bin\_classify,bin\_annotation.

RUN

Reset

0

搜索

20:51  
2023-05-04

**analytical  
methods**





**DATA**



# Machine learning prediction of the degree of food processing

Received: 7 May 2022

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**NOVA** is a classification system widely used in epidemiological studies, assessing the extent and purpose of food processing.

It categorizes individual foods into four broad categories:

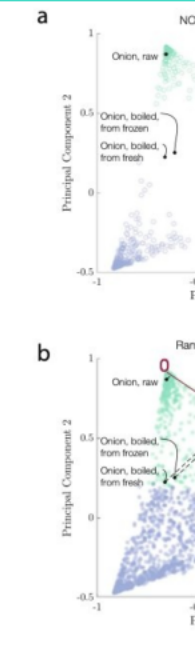
**NOVA 1:** unprocessed or minimally processed, like fresh, dry, or frozen fruits or vegetables, grains, legumes, meat, fish, and milk; processed culinary ingredients

**NOVA 2:**, like table sugars, oils, fats, and salt;

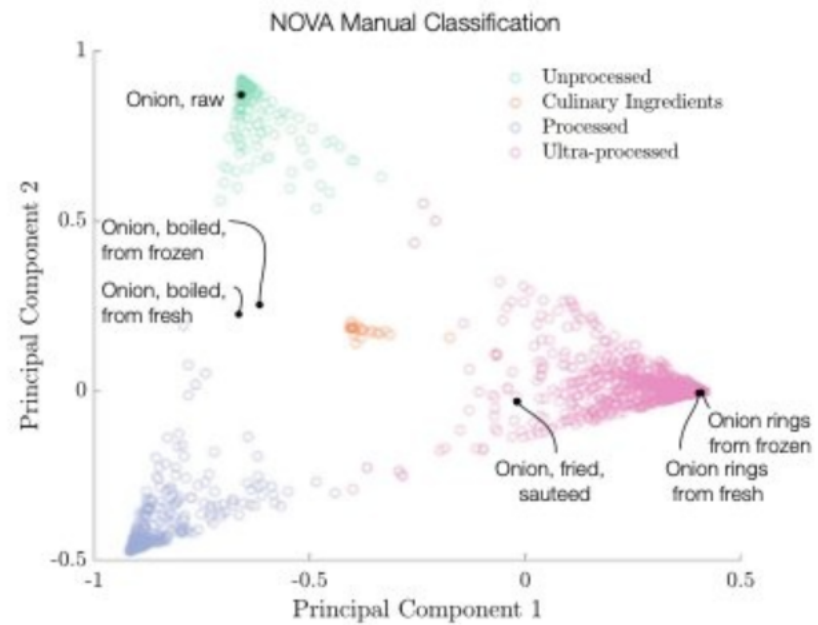
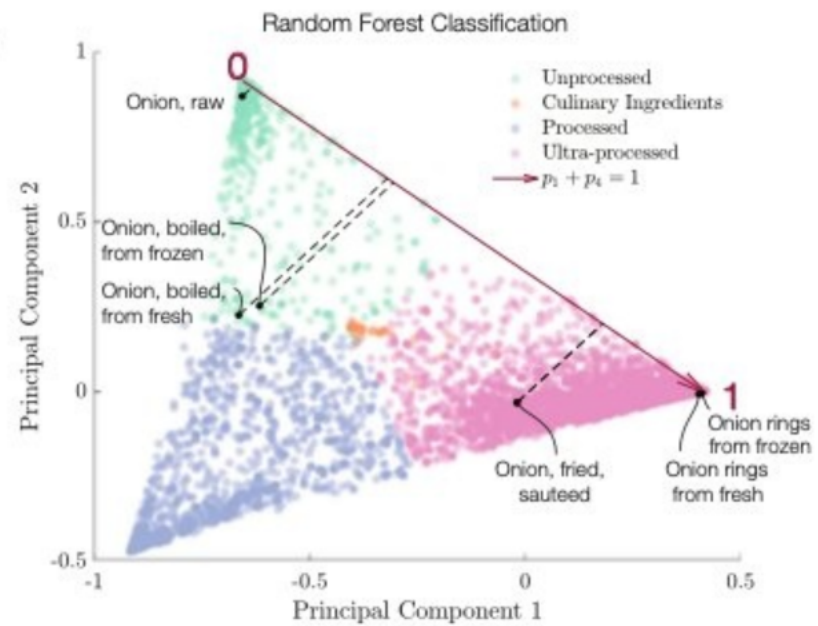
**NOVA 3:**, Processed foods like canned food, simple bread, and cheese; and

**NOVA 4: Ultra processed products** industrial formulations typically of five or more ingredients including substances not commonly used in culinary preparations, such as additives whose purpose is to imitate sensory qualities of fresh food.

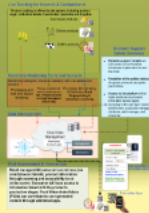
**Epidemiological studies have documented significant associations between greater consumption of NOVA 4 and disease onset, including links to obesity, CHD, diabetes mellitus, cancer, and depression.**



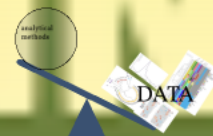


**a****b**

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A diagram titled "INTERNET of FOODS" showing a white cloud with several red Wi-Fi signal icons hanging from it. Below the cloud are various food items: a carton of milk, a glass of milk, a salmon fillet, a wedge of Swiss cheese, and a bunch of corn cobs, illustrating the concept of food connectivity.

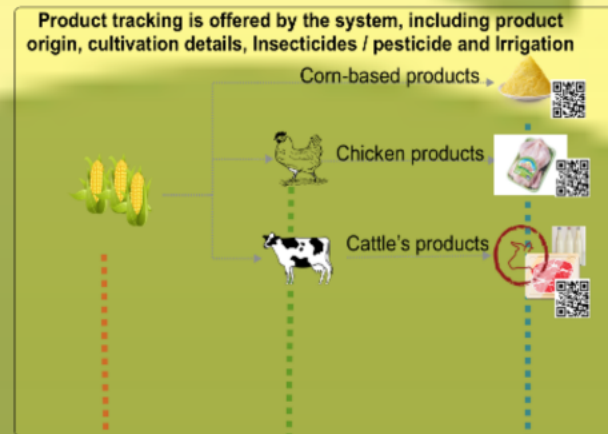


**An explosion of data – on chain**  
The gradual move from paper-based one-up, one-down records to extensive electronic records  
Automatic data capture, via barcodes, RFID, IoT, integrated into ERP systems (e.g. Returnable Transport Items with RFID or even IoT devices)  
**On the farm:** weather stations, sensors in the soil, on livestock, in machinery, from drones and Earth Observation.  
**In the food processors:** ERP systems capturing GSI/EPCIS events, processing, packaging, transforming.  
**In the logistics stack:** GPS tracking, sensors monitoring (e.g. temperature, humidity)

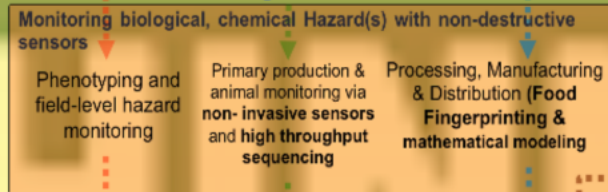
The massive amount of data generated by various analytical and high throughput platforms is a challenging issue for food safety.

[illegible]

## Live Tracking for Hazards & Contaminants



### Real-time Monitoring Tools and Sensors



### Data Management



### Risk Assessment & Intervention

Retail managers/QC personell can retrieve (via smartphones /tablets) product information through scanning and accessibility to an online server. Consumer will have access to information linked with the product's production stages. Food Value-chain Actors (FVAs) can contribute to user-generated content through additional apps.

### Decision Support Safety Services)

- **Decision-support models** are built based on non-invasive techniques implemented across the chain
- **Prediction of the safety indices** of a given product at any given point of time.
- **Access to the platform** will be made available over several information access layers according to the user type: system administrator, production manager, distributor, retail manager, and consumer.



User generated content & Communication Apps

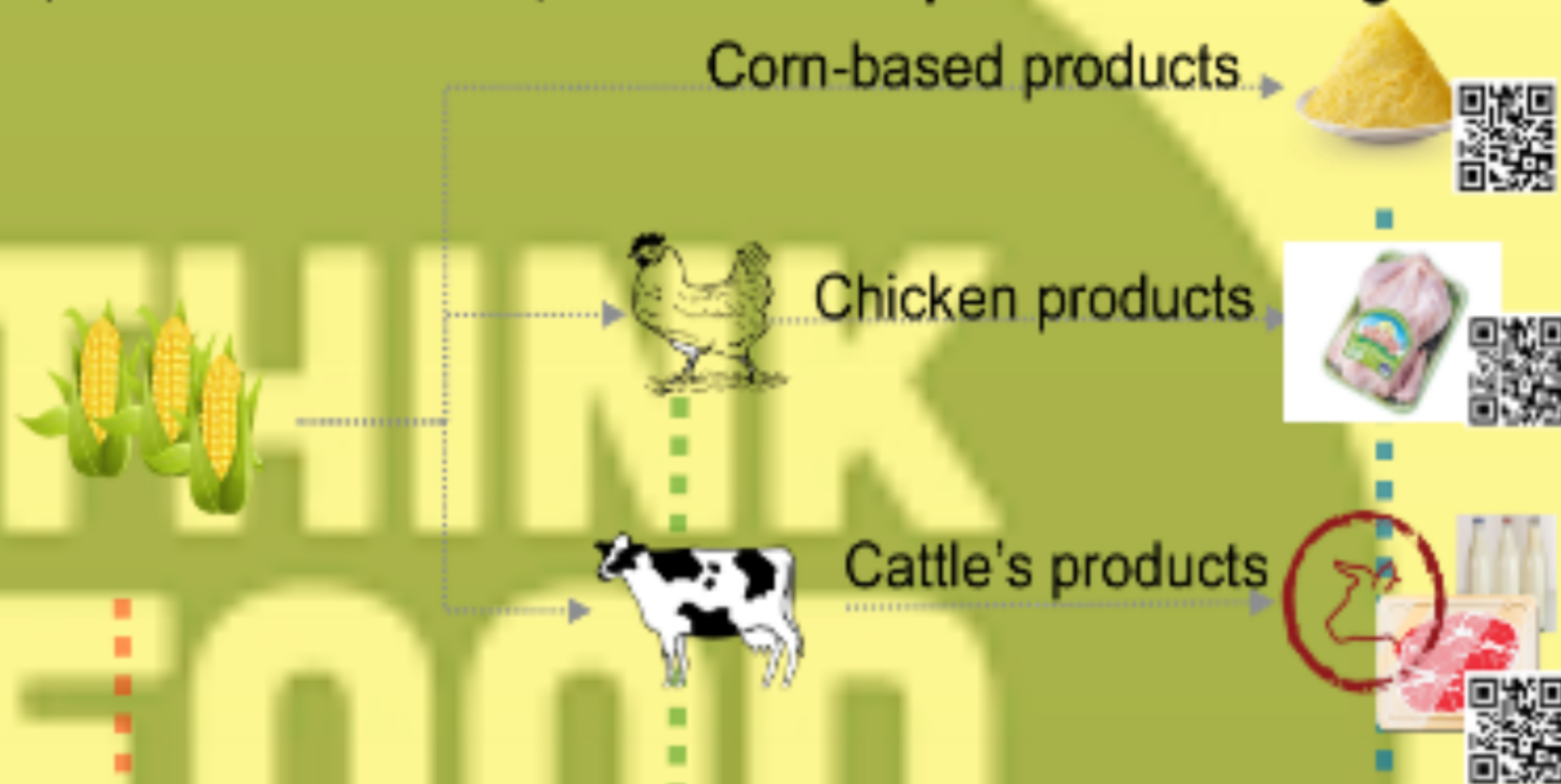
Food safety Apps

Pilot 1 Gr



## Live Tracking for Hazards & Contaminants

Product tracking is offered by the system, including product origin, cultivation details, Insecticides / pesticide and Irrigation



- **Decis**  
built b  
techni



## Real-time Monitoring Tools and Sensors

Monitoring biological, chemical Hazard(s) with non-destructive sensors

Phenotyping and field-level hazard monitoring

Primary production & animal monitoring via non-invasive sensors and high throughput sequencing

Processing, Manufacturing & Distribution (Food Fingerprinting & mathematical modeling)

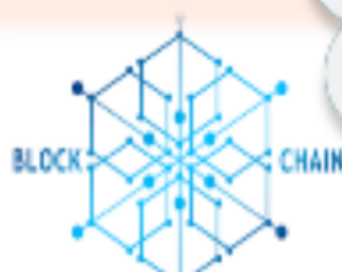
## Data Management

Cloud Data Management

Information Access Layer

Management

Distribution



techniques implemented in the chain

- **Prediction of the** of a given product at a point of time.
- **Access to the platform** made available over information access according to the user administrator, producer, distributor, retail merchant, consumer.

smartphone

Corn & food (meat) products



attle's products



## Decision Support Safety Services)

and Sensors

(s) with non-destructive

Processing, Manufacturing  
& Distribution (**Food  
Fingerprinting &  
mathematical modeling**)

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Management

Distribution

retrieve (via  
information  
to an  
access to  
s  
n Actors  
ated

smartphone

Corn & food (meat)  
products

Product Origin

Reviews ★★★★★

Nutrition Value

Food Miles

Safety Profile

Tracking information

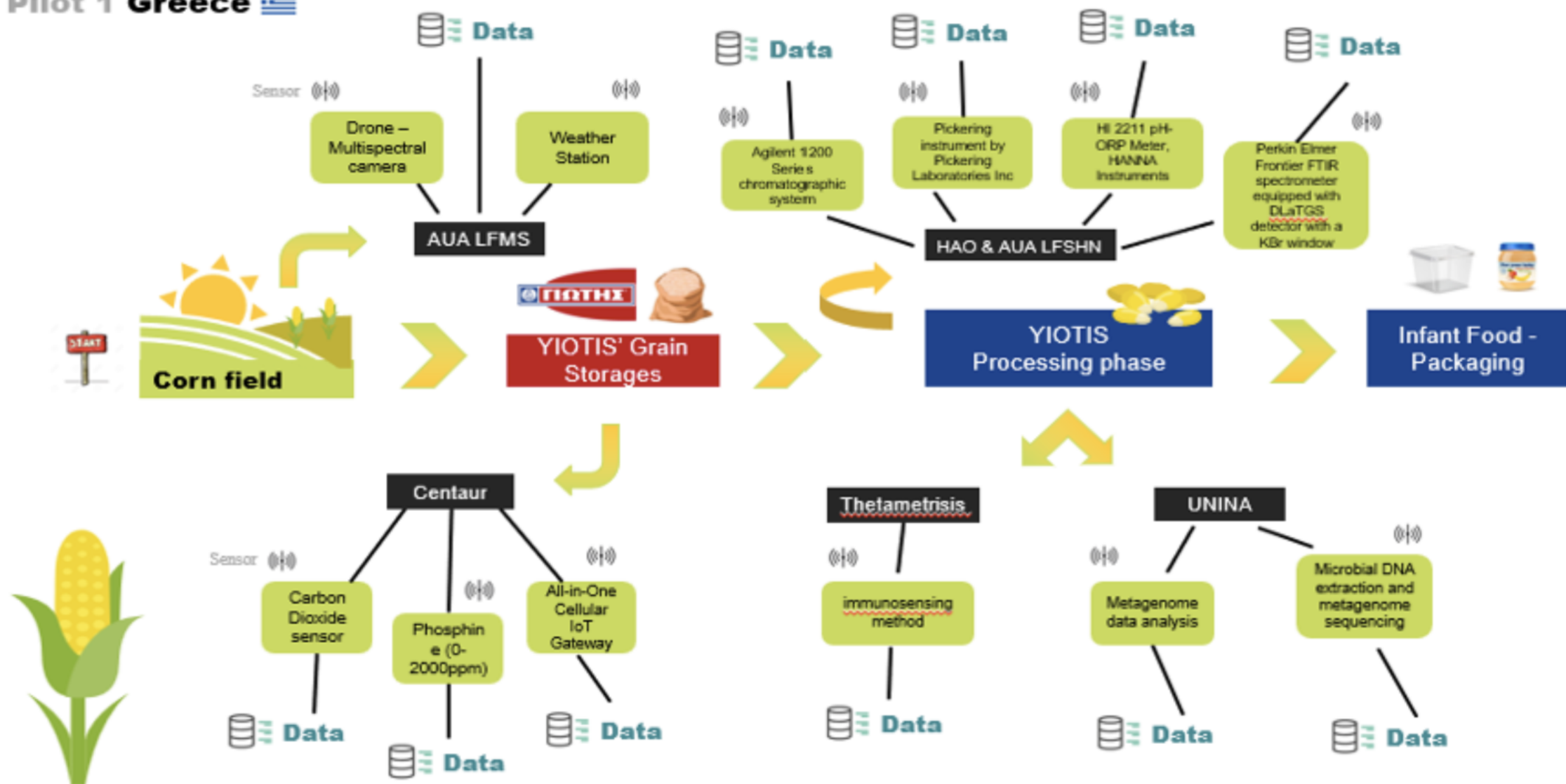
Predictive Hazard e.g.  
(mycotoxin, pathogen)

User generated content  
& Communication Apps

Food safety Apps



## Pilot 1 Greece 🇬🇷





## **An explosion of data – on chain**

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## **AN EXPLOSION OF DATA -OFF CHAIN**

Two areas:

### **1. Food safety data collection**

Traditional methods i.e. sample based microbiological laboratory analysis (“finished product testing”)

More recently non-destructive methods – so called “process analytical technology” (Nychas et al. 2016) (e.g. FTIR spectra)

Lots of mathematical modelling, growing use of machine learning and deep learning (Nychas et al. 2021)

Here we are not concerned with limitations/ capabilities of these approaches only to say they generate a lot data – but where is it? Who has access?

### **2, Food certification:**

or quality features of food (GlobalGAP, organic, Fairtrade, and many other certification systems)

In all these cases data resides with different organisations and institutions but is not available to the community.

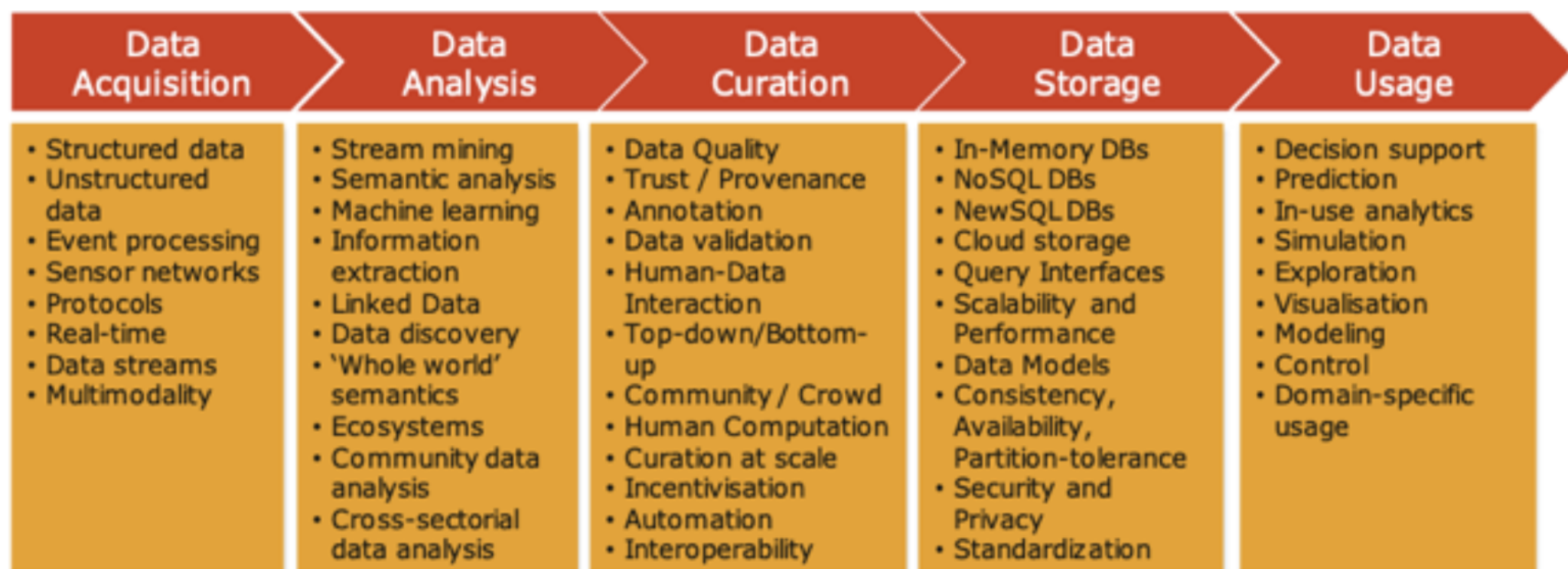
analytical  
methods

DATA



# Data Science; Data mining, Data analysis, Machine Learning

The massive amount of data generated by various analytical and high throughput platforms is a challenging issue for food safety.



**Technical Working Groups**

Cloud platforms and data repositories should be coupled with appropriate web applications in order to assist producers with their expenditure and planning decisions.

## ***INTERNET of FOODS***





Review

# Artificial Intelligence in Food Safety: A Decade Review and Bibliometric Analysis

Zhe Liu <sup>1,\*</sup>, Shuzhe Wang <sup>1</sup>, Yudong Zhang <sup>2,\*</sup>, Yichen Feng <sup>1</sup>, Jiajia Liu <sup>1</sup> and Hengde Zhu <sup>2</sup>

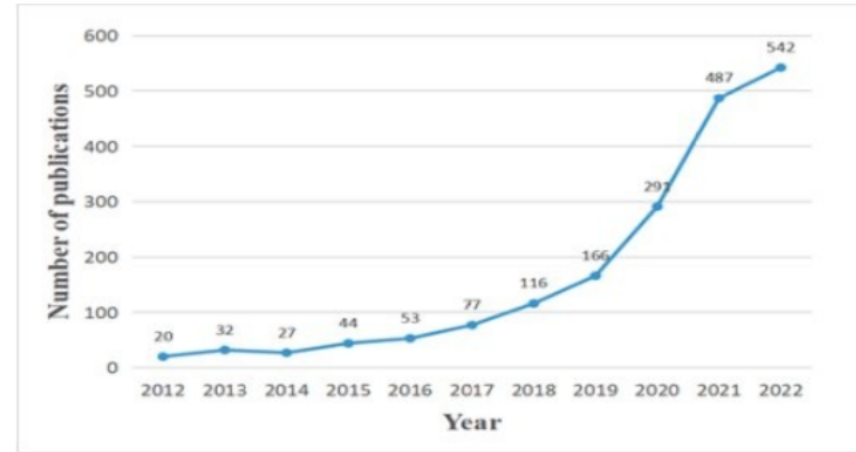


Figure 2. AI-related publications on food safety from 2012 to 2022.

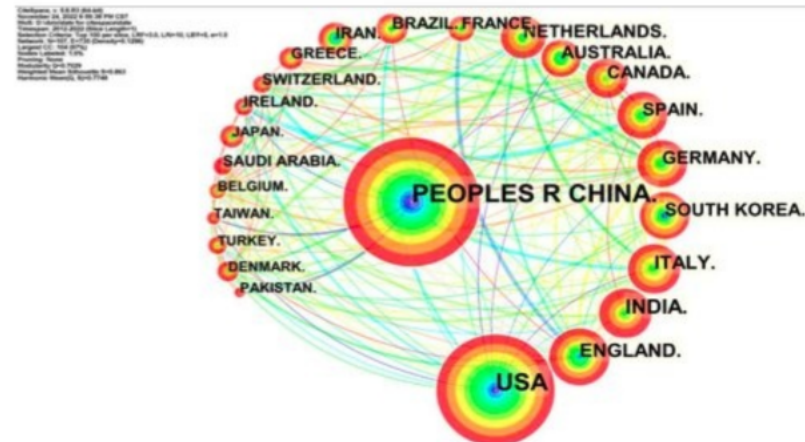


Figure 11. The country co-authorship network of AI-related publications.



**THANK YOU FOR YOUR ATTENTION!**

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# ***Food Safety and Quality under the Auspices of Data Science***

## ***IAFP : Webinar 24/10/2023***



**George-John NYCHAS**  
**Agricultural University of Athens -**  
**Greece**

