



INRAE



BORDEAUX
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Session 89

Optimizing rearing performance through algorithmic approaches to maximize meat quality in livestock

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Headline



Introduction



OptiMeat V.1



Application



Next Step



Introduction



Introduction

Consumer is more concerned

- Animal rearing system
- Sensorial and nutritional quality
- Environmental impact





Introduction

Adapting to consumer preferences and meet
consumer expectations is mandatory for the viability
of the market



How to optimize breeding practices to respond for market demand

Multi-objective algorithm

Deterministic

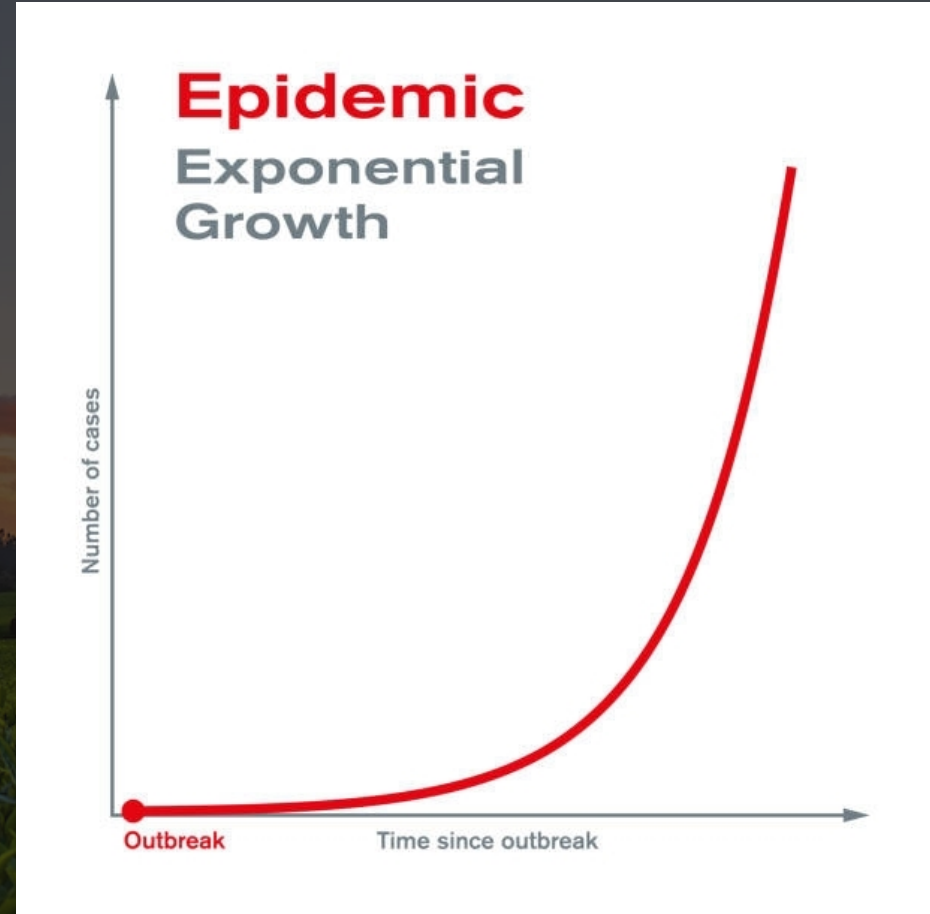
| Object | Combinaison | Temps |
|--------|-------------|-----------|
| 5 | 32 | 0.0003 s |
| 10 | 1024 | 0.0008 s |
| 20 | 1 048 576 | 1 s |
| ... | | |
| 77 | 23 722 485 | 5000 year |

100%

Stochastic

| Objet | Combinaison | Temps |
|-------|-------------|-------|
| 77 | 23 722 485 | 1.7 s |

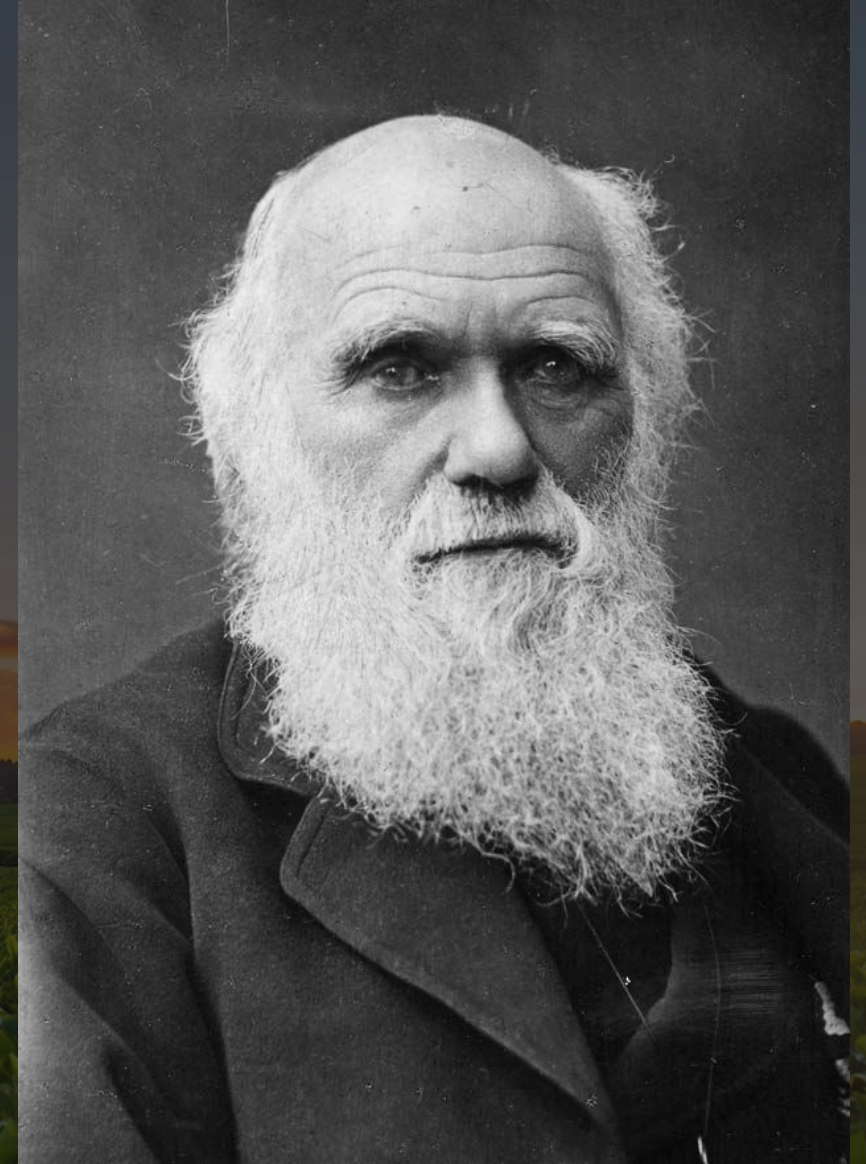
93%

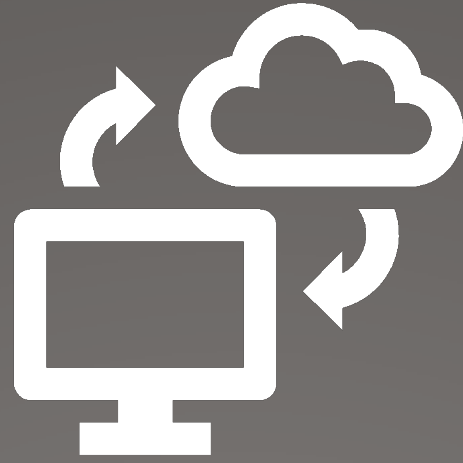


Genetic Algorithm

Stochastic Optimization
Methods
Darwin's theory

Often used to find the best solution to a problem, when there is no deterministic method to calculate the exact solution.





OptiMeat V.1

OptiMeat v.1

INPUT

Decision Matrix

Body weight
Metabolizable Energy
Ageing
Feed consumption

Objective Matrix

Maximization Minimization

Tenderness
Juiciness
Flavor
Lipid



OptiMeat v.1



| Decision Matrix | | | Objective Matrix | |
|-----------------|----|--------|-----------------------------------|-------------------------------|
| ME | BW | Ageing | Tenderness <i>Maximization</i> | Flavor <i>Maximization</i> |
| 10 | 2 | 3 | 4.3 | 12.4 |
| 12 | 3 | 5 | 6.3 | 20.4 |
| 11 | 7 | 6 | 14.3 | 24.4 |

OptiMeat v.1



$$\text{Tenderness} = 2 + 0.3 \text{ BW}$$

$$\text{Flavor} = 4 + 0.4 \text{ Ageing} + 0.2 \text{ ME}$$

| Decision Matrix | | | Objective Matrix | |
|-----------------|----|--------|------------------|--------|
| ME | BW | Ageing | Tenderness | Flavor |
| 10 | 2 | 3 | 4.3 | 12.4 |
| 12 | 3 | 5 | 6.3 | 20.4 |
| 11 | 7 | 6 | 14.3 | 24.4 |

OptiMeat v.1



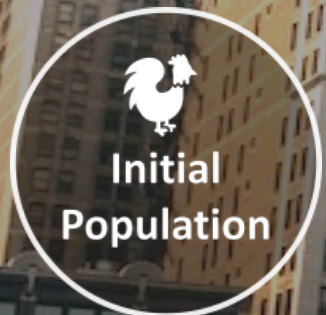
| Decision Matrix | | |
|-----------------|----|--------|
| ME | BW | Ageing |
| 10 | 2 | 3 |
| 12 | 3 | 5 |
| 11 | 7 | 6 |

Fitness

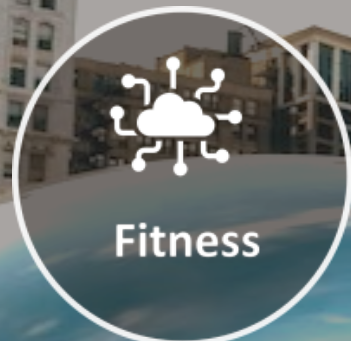
$$\text{Tenderness} = 2 + 0.3 \text{ BW}$$

$$\text{Flavor} = 4 + 0.4 \text{ Maturatin} + 0.2 \text{ ME}$$

OptiMeat v.1



Decision Matrix



$$f(X) = a_1x_1 + a_2x_2 + a_3x_3$$





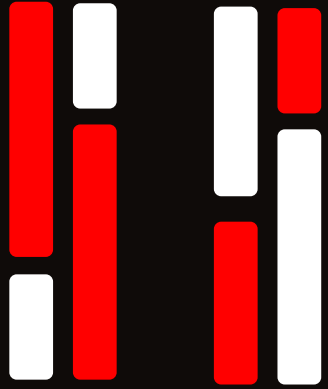
**Genetic
operator**



Genetic operator

Generation: n times





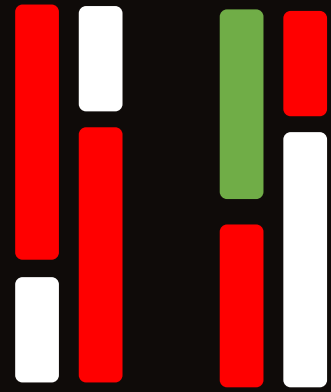
CrossOver



Crossover

(Bloc Swap)

| Individual | ME | BW | Ageing |
|------------|----|-----|--------|
| Parent 1 | 42 | 500 | 5 |
| Parent 2 | 50 | 300 | 15 |
| Child 1 | 42 | 500 | 15 |
| Child 2 | 50 | 300 | 5 |



Mutation

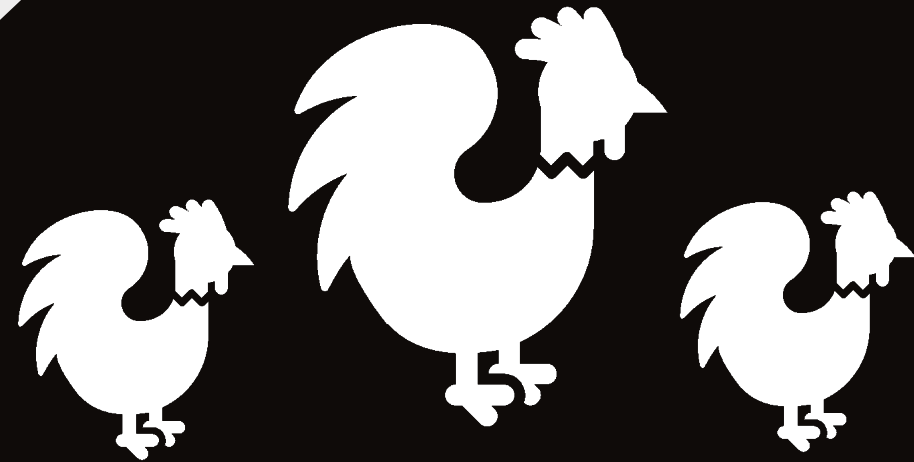


Mutation

Maximum-Minimum

| | minimum | maximum |
|----|---------|---------|
| BW | 200 | 400 |

| Individual | ME | BW | Ageing |
|------------|----|-----|--------|
| Parent 1 | 42 | 500 | 5 |
| Child 2 | 42 | 300 | 5 |



Children
= N size



Population

Parent = Decision Matrix

| ME | BW | Ageing |
|----|----|--------|
| 10 | 2 | 3 |
| 12 | 3 | 5 |
| 11 | 7 | 6 |

Children = Decision Matrix

| ME | BW | Ageing |
|----|-----|--------|
| 42 | 500 | 15 |
| 50 | 300 | 5 |
| 50 | 300 | 5 |



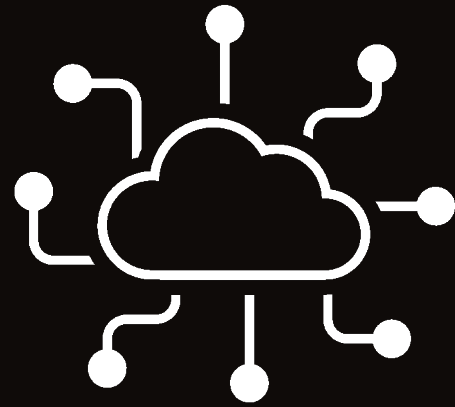
Population

Decision Matrix

Parent

Child

| | ME | BW | Ageing |
|--------|----|-----|--------|
| Parent | 10 | 2 | 3 |
| | 12 | 3 | 5 |
| | 11 | 7 | 6 |
| Child | 42 | 500 | 15 |
| | 50 | 300 | 5 |
| | 50 | 300 | 5 |



Fitness



Fitness

Decision Matrix

| | ME | BW | Ageing |
|---------------|-----------|-----------|---------------|
| Parent | 10 | 2 | 3 |
| | 12 | 3 | 5 |
| | 11 | 7 | 6 |
| Child | 42 | 500 | 15 |
| | 50 | 300 | 5 |
| | 50 | 300 | 5 |

Parent

Child

Objective Matrix

| Tenderness | Flavor |
|-------------------|---------------|
| 4.3 | 12.4 |
| 6.3 | 20.4 |
| 14.3 | 24.4 |
| ? | ? |
| ? | ? |
| ? | ? |

Parent

Child



Fitness

$$\text{Tenderness} = 2 + 0.3 \text{ BW}$$

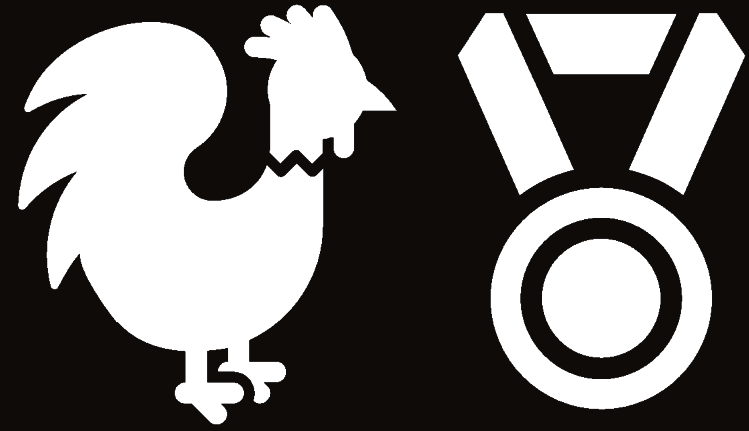
$$\text{Flavor} = 4 + 0.4 \text{ Maturatin} + 0.2 \text{ ME}$$

Decision Matrix

| | ME | BW | Ageing |
|--------|----|-----|--------|
| Parent | 10 | 2 | 3 |
| Parent | 12 | 3 | 5 |
| Parent | 11 | 7 | 6 |
| Child | 42 | 500 | 15 |
| Child | 50 | 300 | 5 |
| Child | 50 | 300 | 5 |

Objective Matrix

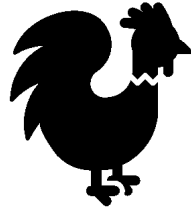
| Tenderness | Flavor |
|------------|--------|
| 4.3 | 12.4 |
| 6.3 | 20.4 |
| 14.3 | 24.4 |
| 4.3 | 12.4 |
| 6.3 | 20.4 |
| 14.3 | 24.4 |



Selection



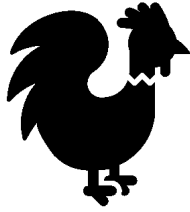
(Pareto Front)
Selection



| Objective Matrix | | |
|------------------|------------|--------|
| | Tenderness | Flavor |
| A | 4.3 | 12.4 |
| B | 6.3 | 20.4 |
| C | 14.3 | 24.4 |
| D | 14.3 | 24.4 |
| E | 4.3 | 12.4 |
| F | 6.3 | 20.4 |
| G | 14.3 | 24.4 |



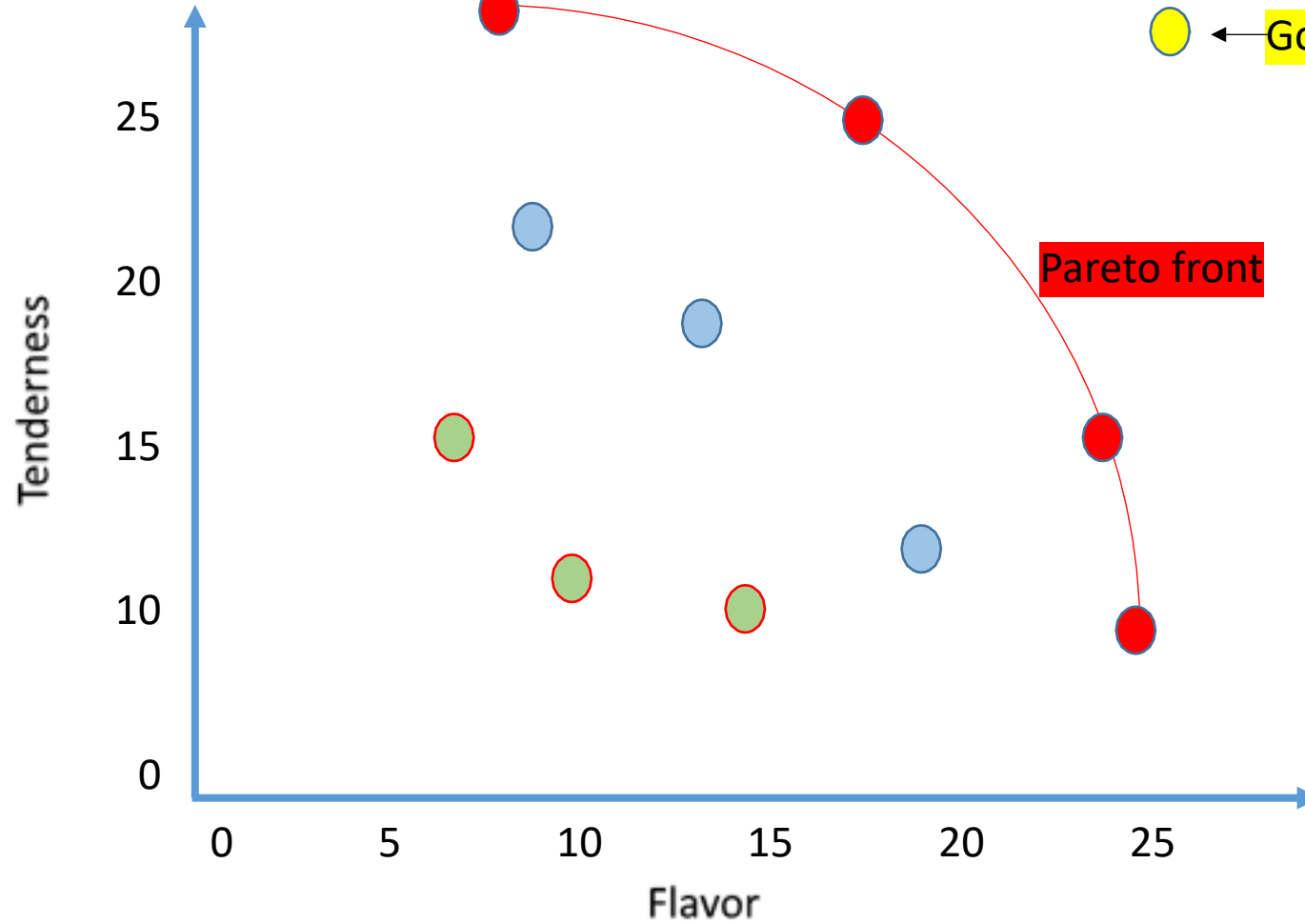
(Pareto Front)
Selection



| ME | BW | Ageing |
|----|----|--------|
| 10 | 2 | 3 |



Golden solution





Headline



Introduction



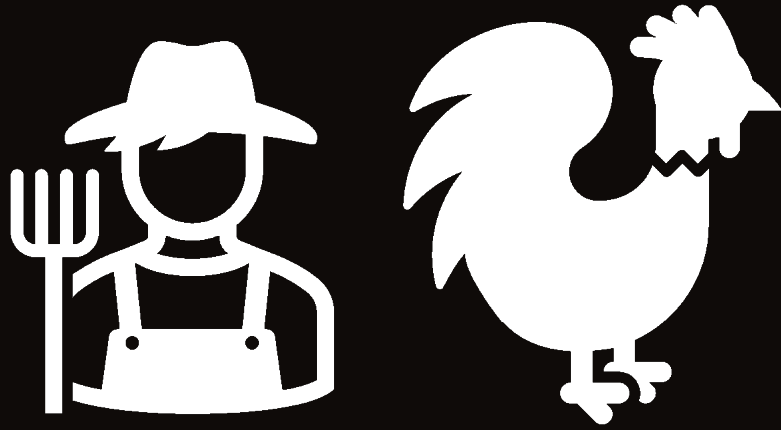
OptiMeat V.1



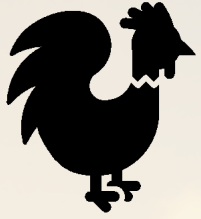
Application



Next Step



Application



Application



ITAVI

INRAE

Label Rouge, Certified, Standard and Heavy Broilers

four main French production systems

7,843 Chicken Breast

Decision Matrix

BW (kg)

Age (day)

Carcass Holding Period (hour)

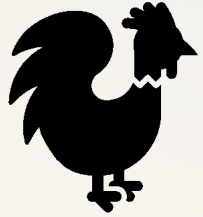
Objective Matrix

Minimization

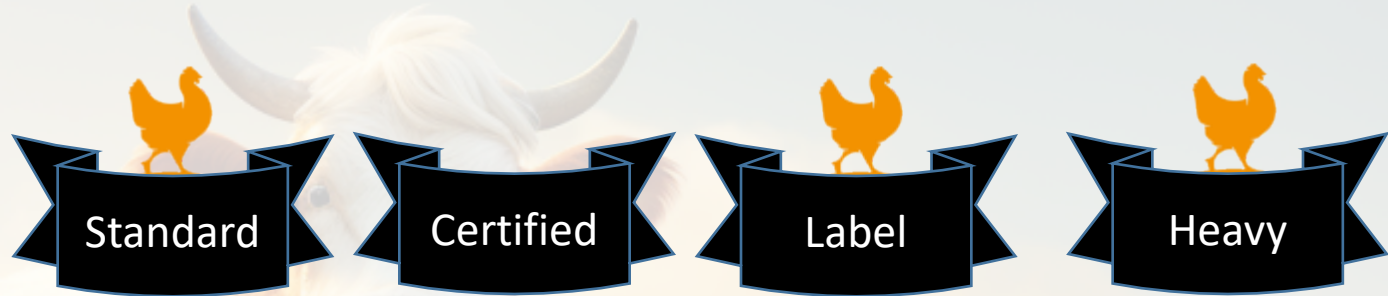
Lipids (%)

Toughness (N)

Cooking loss (%)



Result



| | Standard | Certified | Label | Heavy | |
|--------------------|---|-----------|-------|-------|-------|
| <i>Generation</i> | 11 | 6 | 8 | 8 | |
| Breeding practices | <i>BW (kg)</i> | 2.39 | 1.80 | 2.42 | 4.78 |
| | <i>Age (day)</i> | 35 | 44 | 92 | N/A |
| | <i>1. Carcass Holding Period (hour) DEBONING TIME PLUS EXPL</i> | 30 | 48 | 8 | 3 |
| | <i>Lipids (%)</i> | 1.62 | 1.19 | 0.95 | 1.70 |
| Objective | <i>Toughness (N)</i> | 9.50 | 17.15 | 20.02 | 15.04 |
| | <i>Cooking loss (%)</i> | 10.65 | 12.36 | 12.44 | 12.77 |



Headline



Introduction



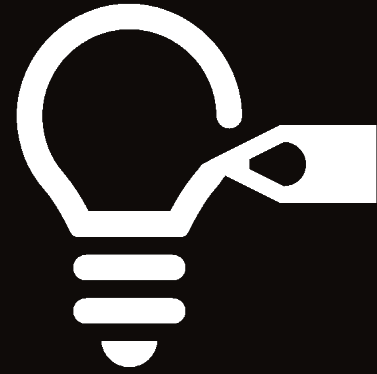
OptiMeat V.1



Application



Next Step



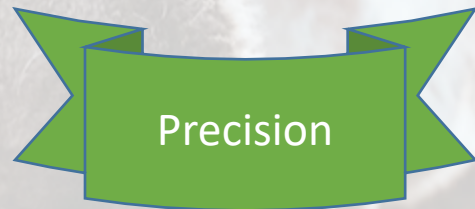
Next Step



Next Step

Evaluation

Nonparametric method
Random forest



Decision Matrix

Implementing
Qualitative variable



Thank you

Acknowledge dgments



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Hocquette**



SCAN ME