











### Assessing the impact of extensive husbandry conditions on broiler meat quality using machine learning



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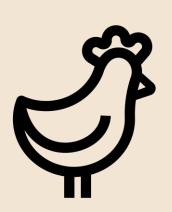


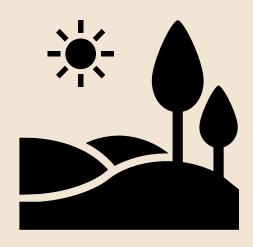






"Linking extensive husbandry practices to the intrinsic quality of broiler meat"















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Space allowance	<b>Low density</b> (≤ 35 kg/m2)	<b>High density</b> (≥ 38 kg/m2)	
Diet	Roughage	No added roughage	
Genetics	Slow-growing, male-layer, dual- purpose	Conventional fast- growing	
Quality of Space	Enrichment (barrier, perch, straw/lucerne bale, dust bath)	No added enrichment	
Welfare	High	Low	





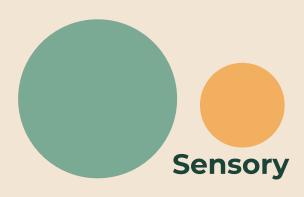


### **Husbandry practices**

- Space allowance
- Diet
- Genetics
- Quality of Space

### Welfare score

- Cleanliness
- Gait score
- Hock burn
- Skin condition



# Intrinsic meat quality



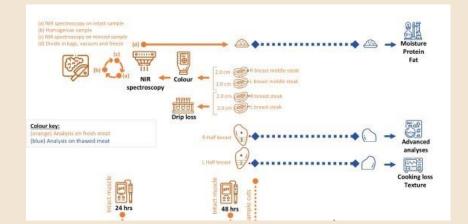
### Chemical



 Trained sensory panel (8 panelists)



 Appearance, odour, flavour, texture and aftertaste  pH, texture analysis, L\*a\*b\* values, moisture content...









## **Data processing**

80% train set (with cross-validation)

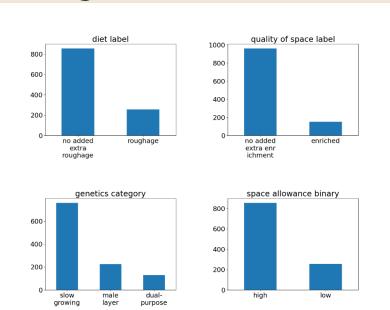
20% test set

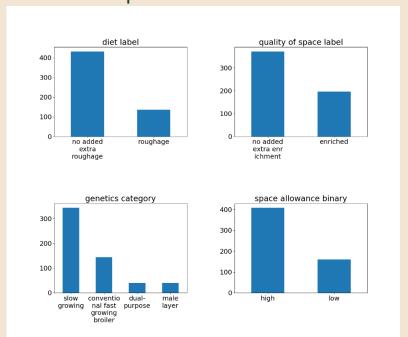
## Chemical

- Min-max scaling
- MICE imputation

### Sensory

Panelist-wise z-score scaling





**mEAT** 

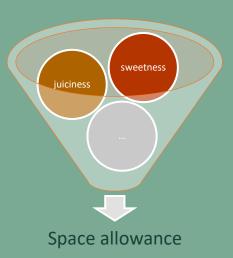




Machine learning (ML) methodology

Can we trace back the extensiveness of a condition from the quality of meat?

- 1) Train ML models to classify husbandry conditions and welfare
- 2) Performance assessment (ROC-AUC)
- 3) Feature importance analysis







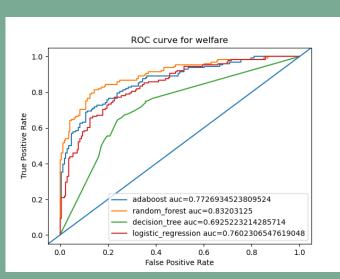


# Performance assessment quality

**Architecture space:** 

AdaBoost, Random Forest, Logistic Regression, Decision Tree

**Best results:** 



	Sensory		Chemical	
	ROC-AUC	Model	ROC-AUC	Model
Genetics	0.77	Random Forest	0.94	AdaBoost
Welfare	0.83	Random Forest	0.92	Random Forest







# Feature importance analysis





# SHAP value-based explanations

### **Genetics:**

sensory: crispiness skin, colour skin, and saltiness skin chemical: colour (a\*, b\*) and texture (BMORS shear energy)

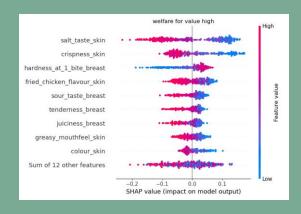
### Welfare:

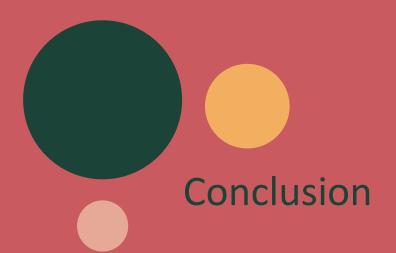
sensory: saltiness of skin, crispiness of skin and hardness at first bite

chemical: colour (a\*b\*), cooking loss, and texture(BMORS shear energy)











**Genetics** and **welfare** have the strongest link to meat quality





### Upcoming...





- Publication of "Assessing the impact of husbandry factors on poultry meat quality using a machine learning approach"
- Data available at request\*













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