



Predicting carcass cut yields in cattle from digital images using artificial intelligence

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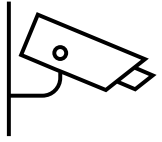


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Bia agus Mara
Department of Agriculture,
Food and the Marine

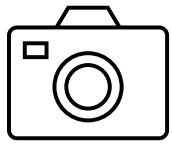


AgTech - it's in our DNA





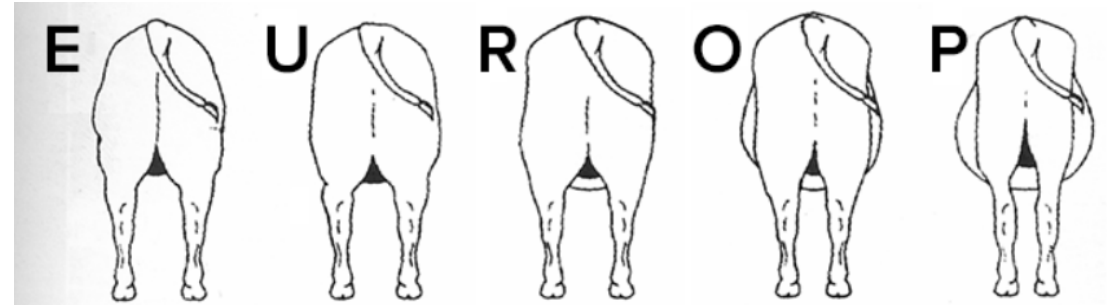
Automated in 2003 using Video Imaging Analysis (VIA)



Captures photo



428 VIA variables measuring carcass dimensions, contours and color



E-
2+

P+
3=



Department of
**Agriculture,
Food and
the Marine**

Conformation score
Fat score
Carcass weight
Slaughter date

DOB, Pedigree, Breed,
etc.



~ 70k Images

Slaney
FOODS
The Beef Specialists



icbof



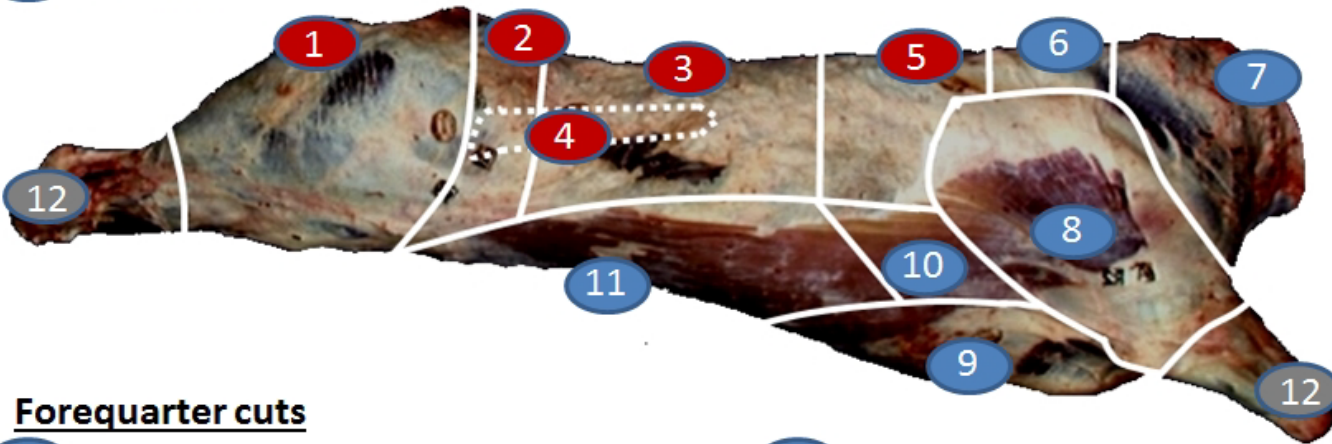
Objective

Predict Cut Yields from Carcass Images using Artificial Intelligence

Cut Groups

Hindquarter cuts

- 1 Round: Silverside, Topside, Knuckle
- 2 Rump
- 3 Striploin
- 4 Fillet
- 5 Cube-Roll



Forequarter cuts

- 6 Chuck
- 7 Neck
- 8 Fore-limb: blade, chuck tender, LMC
- 9 Brisket
- 10 Middle ribs
- 11 Flank & Bavette

Other cuts



- 12 Heel & Shank
- Trimmings

Roasting = Topside + Silverside
+ Knuckle + Rump

Grilling = Striploin + Fillet
+ Cube-roll

1. Phenotypic Data 

2. Carcass Images 

3. Phenotypic Data + Carcass Image Measurements  

=

=

=



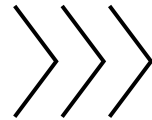
Roasting
Grilling

1. Phenotypic Data



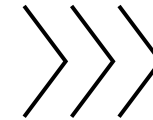
Phenotypic Data

Conformation score
Fat score
Carcass weight
Age at Slaughter
Breed
Slaughter month
Animal type



13 Regression ML algorithms

Linear
Ridge
SGD
Elastic Net
Lars
Lasso
Bayesian Ridge
kNN
Decision Tree
Random Forest
SVR
Gradient Boosting
AdaBoost

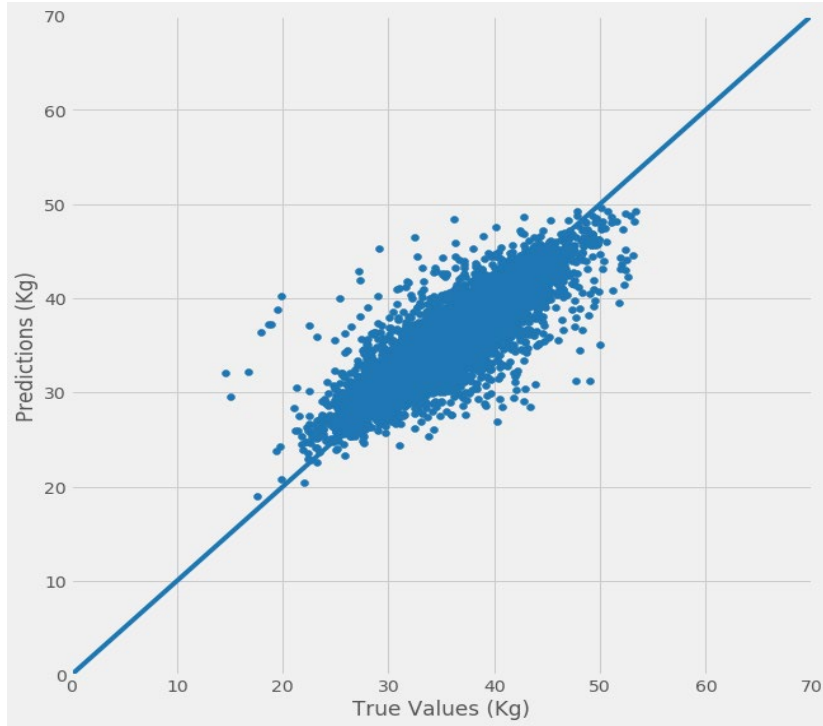


3 Best Performing Models

Hyper-parameter Optimisation
using GridSearchCV

Ridge
Bayesian Ridge
Gradient Boosting

1. Phenotypic Data Results

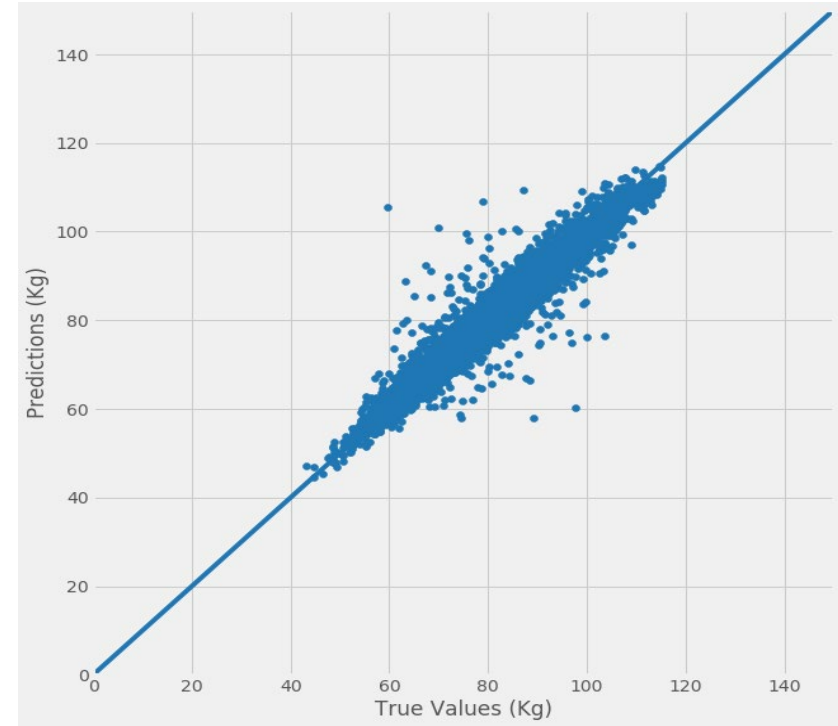


Grilling Cuts

RMSE: 2.94 kg

RRMSE: 8.4%

R²: 0.70



Roasting Cuts

RMSE: 3.23 kg

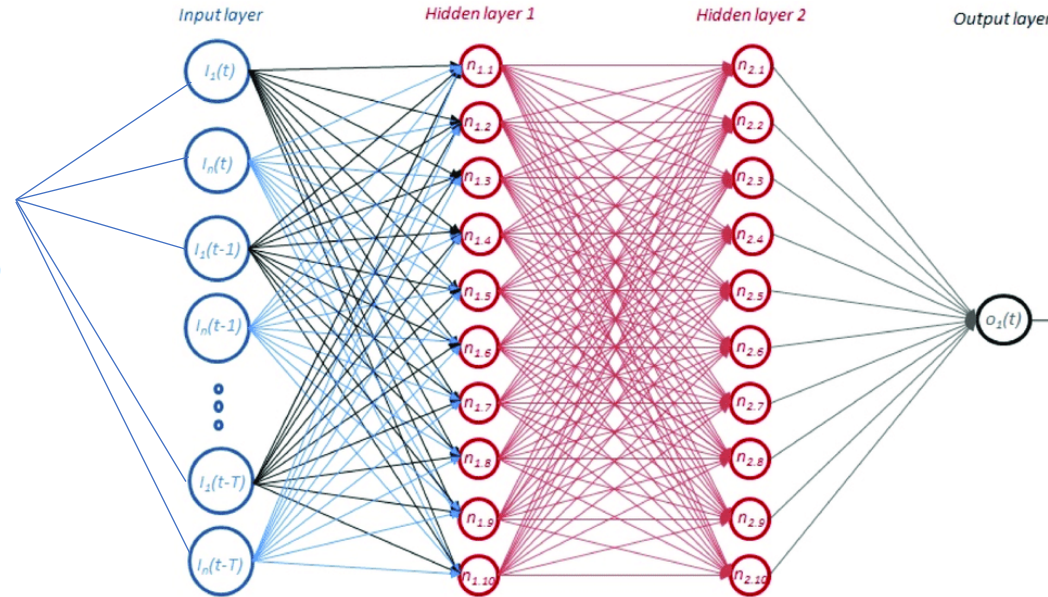
RRMSE: 4.1%

R²: 0.93

2. Carcass Images



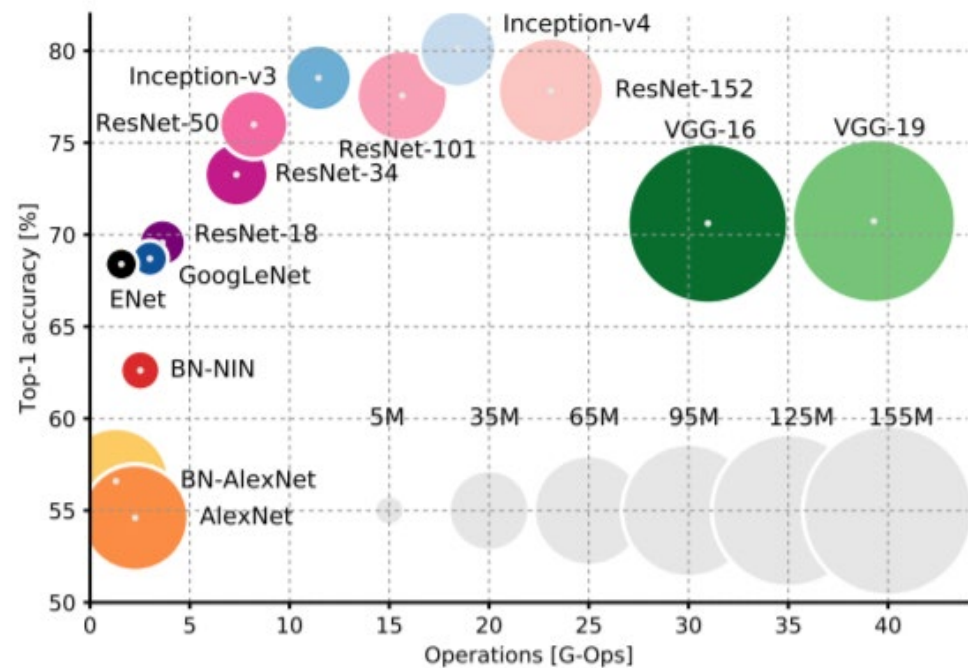
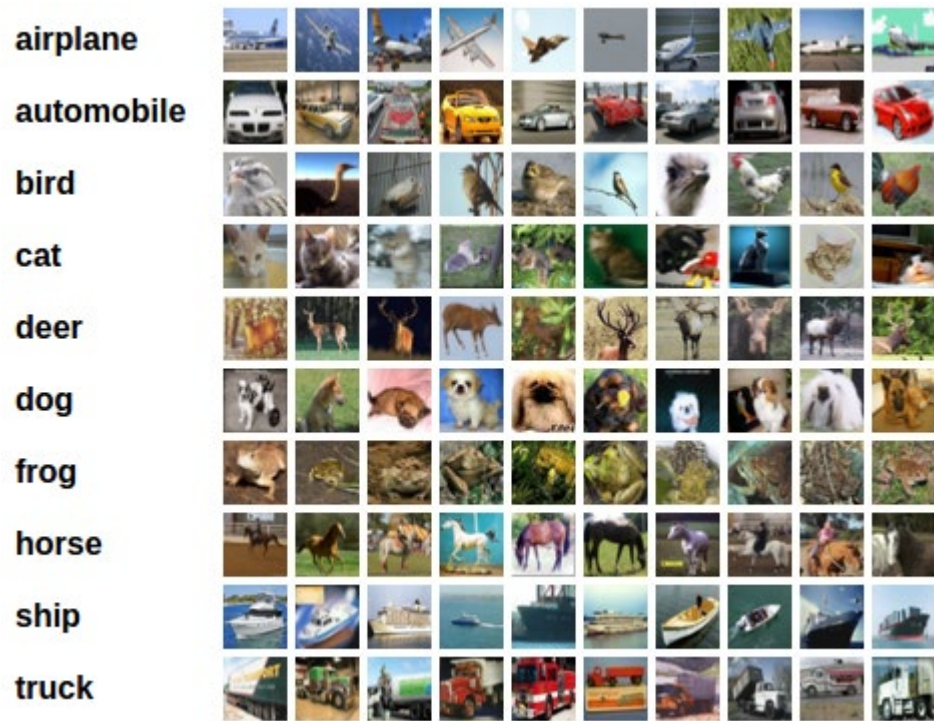
```
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       [   4,   8,  16]],  
       [[ 49, 42, 31],  
        [ 38, 37, 30],  
        [ 43, 35, 21],  
        ...,  
        [ 12(t-T),  
         1n(t-T)]])
```



CNN Architecture Decisions!!



Variable	No. of Options
No. of Hidden Layers	∞
No. of neurons/kernels in every layer	∞
Filter size	1 to Image size
Pool size	1 to Image size
Stride length	1 to Image size
Padding	2



ImageNet

- >14 M labeled images
- >20 k categories
- >500 images per category

ImageNet Visual Recognition Challenge

- Annual since 2010
- CNN's have won since 2012
- Model's are open source

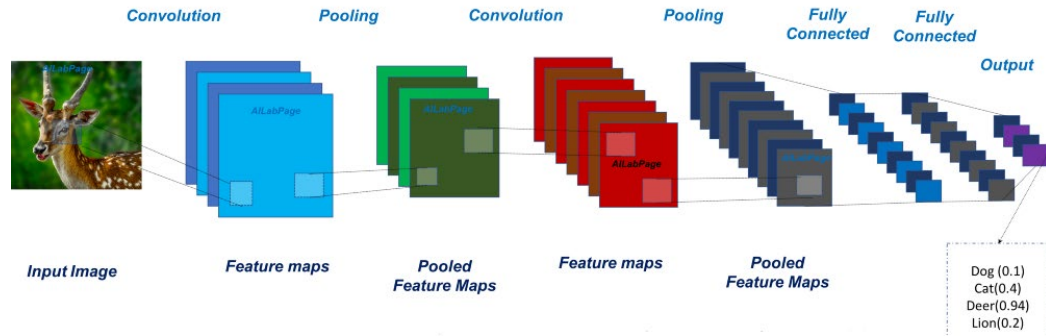
Graph from A. Canziani et al. 2017. An analysis of deep neural network models for practical applications

Pre-trained Models

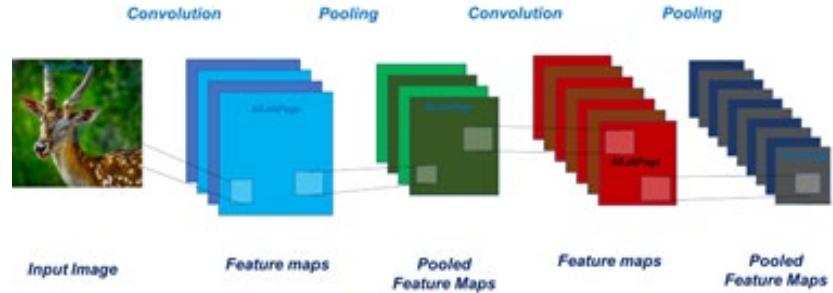
- Deep Learning models available with pre-trained weights
- Can be used for prediction, feature extraction or transfer learning

Model	Size	Top-1 Accuracy	Top-5 Accuracy	Parameters	Depth
Xception	88 MB	0.790	0.945	22,910,480	126
VGG16	528 MB	0.713	0.901	138,357,544	23
VGG19	549 MB	0.713	0.900	143,667,240	26
ResNet50	98 MB	0.749	0.921	25,636,712	-
ResNet101	171 MB	0.764	0.928	44,707,176	-
ResNet152	232 MB	0.766	0.931	60,419,944	-
ResNet50V2	98 MB	0.760	0.930	25,613,800	-
ResNet101V2	171 MB	0.772	0.938	44,675,560	-
ResNet152V2	232 MB	0.780	0.942	60,380,648	-
InceptionV3	92 MB	0.779	0.937	23,851,784	159
InceptionResNetV2	215 MB	0.803	0.953	55,873,736	572
MobileNet	16 MB	0.704	0.895	4,253,864	88
MobileNetV2	14 MB	0.713	0.901	3,538,984	88
DenseNet121	33 MB	0.750	0.923	8,062,504	121
DenseNet169	57 MB	0.762	0.932	14,307,880	169
DenseNet201	80 MB	0.773	0.936	20,242,984	201
NASNetMobile	23 MB	0.744	0.919	5,326,716	-
NASNetLarge	343 MB	0.825	0.960	88,949,818	-
EfficientNetB0	29 MB	-	-	5,330,571	-
EfficientNetB1	31 MB	-	-	7,856,239	-
EfficientNetB2	36 MB	-	-	9,177,569	-
EfficientNetB3	48 MB	-	-	12,320,535	-
EfficientNetB4	75 MB	-	-	19,466,823	-
EfficientNetB5	118 MB	-	-	30,562,527	-
EfficientNetB6	166 MB	-	-	43,265,143	-
EfficientNetB7	256 MB	-	-	66,658,687	-

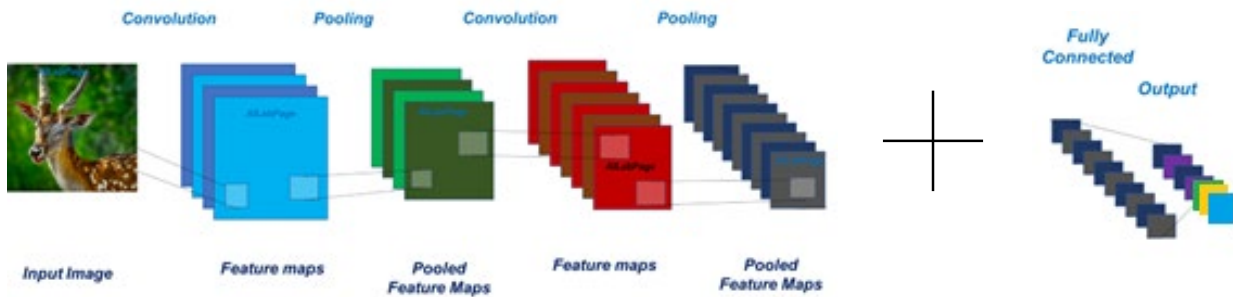
Transfer Learning



Pre-Trained CNN



Remove Top Layers

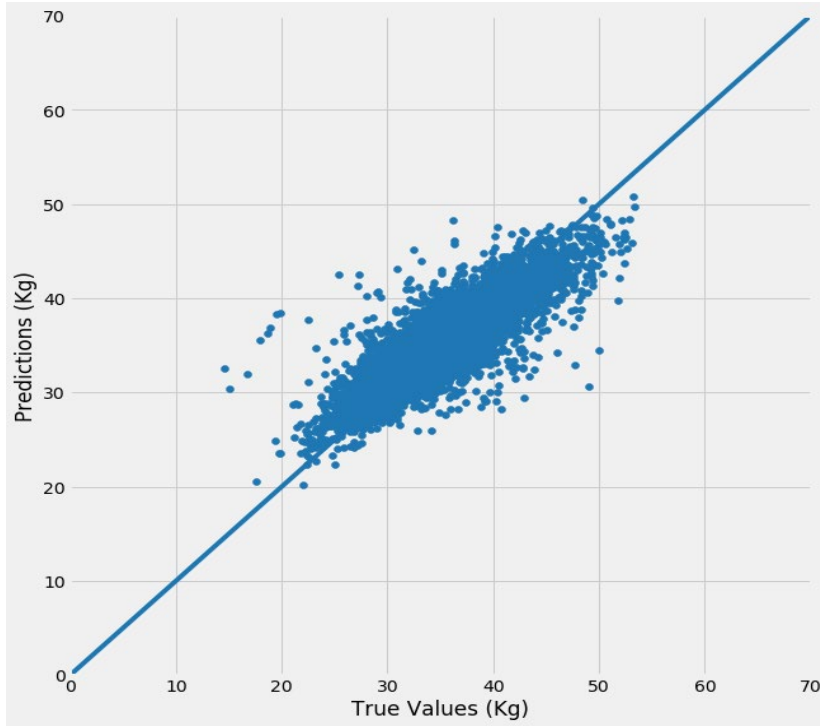


Add Problem Specific Top Layers



Retrain on Carcass Images

2. Carcass Images Results



Grilling Cuts

RMSE: 2.84 kg

RRMSE: 8.1%

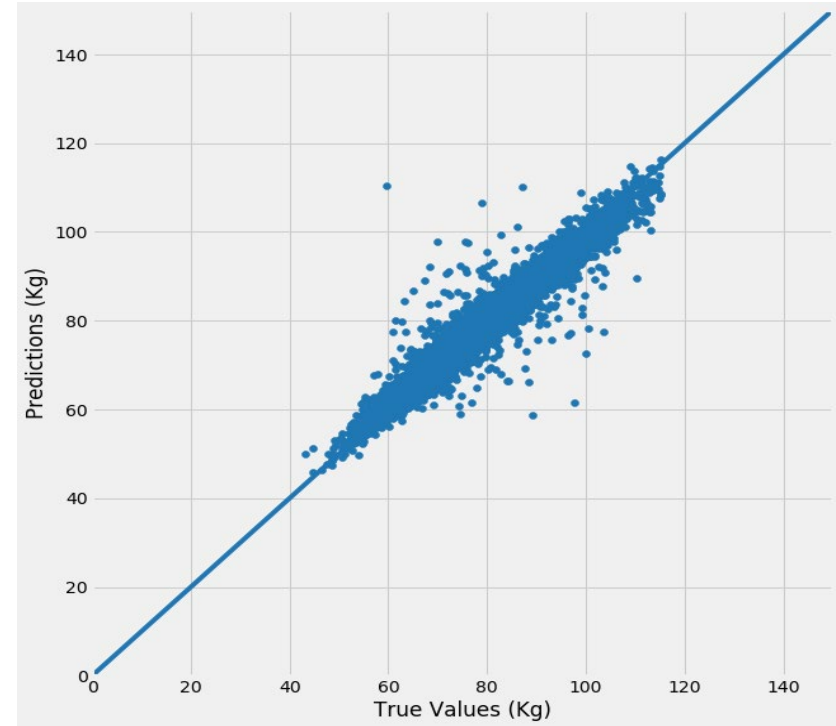
R²: 0.72

1. Phenotypic

vs 2.94 kg

vs 8.4%

0.70



Roasting Cuts

RMSE: 3.26 kg

RRMSE: 4.2%

R²: 0.93

1. Phenotypic

vs 3.23 kg

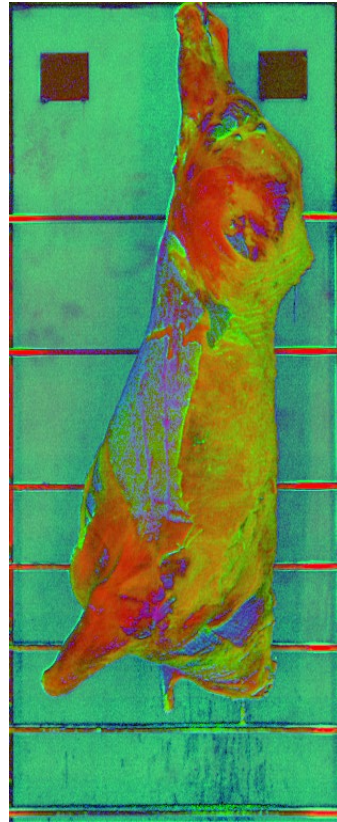
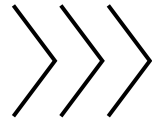
vs 4.1%

0.93

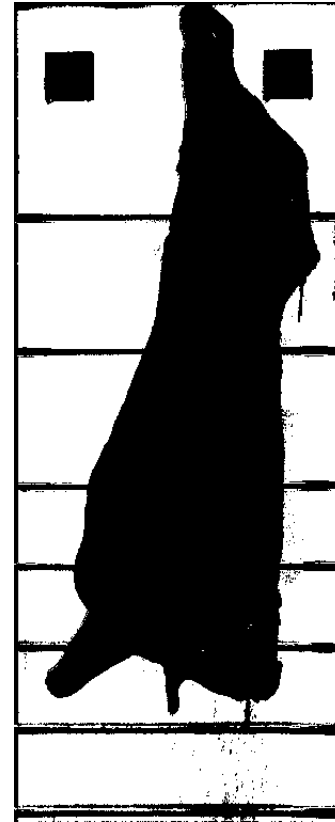
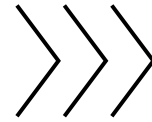
3. Phenotypic Data + Carcass Image Measurements



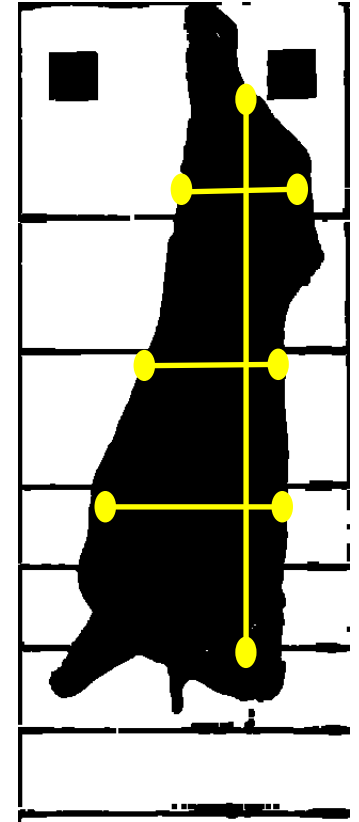
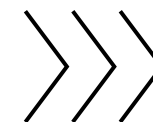
RGB



HSV



Mask



Transform

> 340 measurements

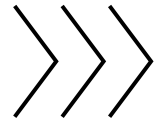
3. Carcass Image Measurements

Carcass Measurements

+

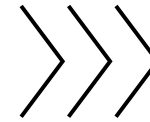
Phenotypic Data

Conformation score
Fat score
Carcass weight
Age at Slaughter
Breed
Animal purpose
Slaughter month
Animal type



13 Regression ML algorithms

Linear
Ridge
SGD
Elastic Net
Lars
Lasso
Bayesian Ridge
kNN
Decision Tree
Random Forest
SVR
Gradient Boosting
AdaBoost



3 Best Performing Models

Hyper-parameter Optimisation
using GridSearchCV

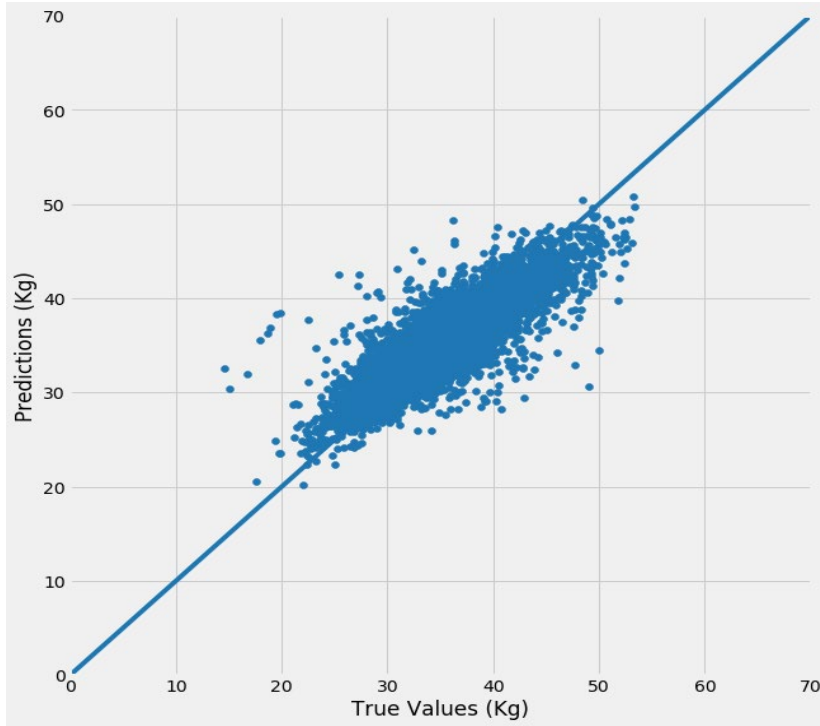
Grill

Random Forest
Bayesian Ridge
Gradient Boosting

Roast

Ridge
Bayesian Ridge
Gradient Boosting

3. Carcass Measurement Results



Grilling Cuts

RMSE: 2.78 kg

RRMSE: 7.9%

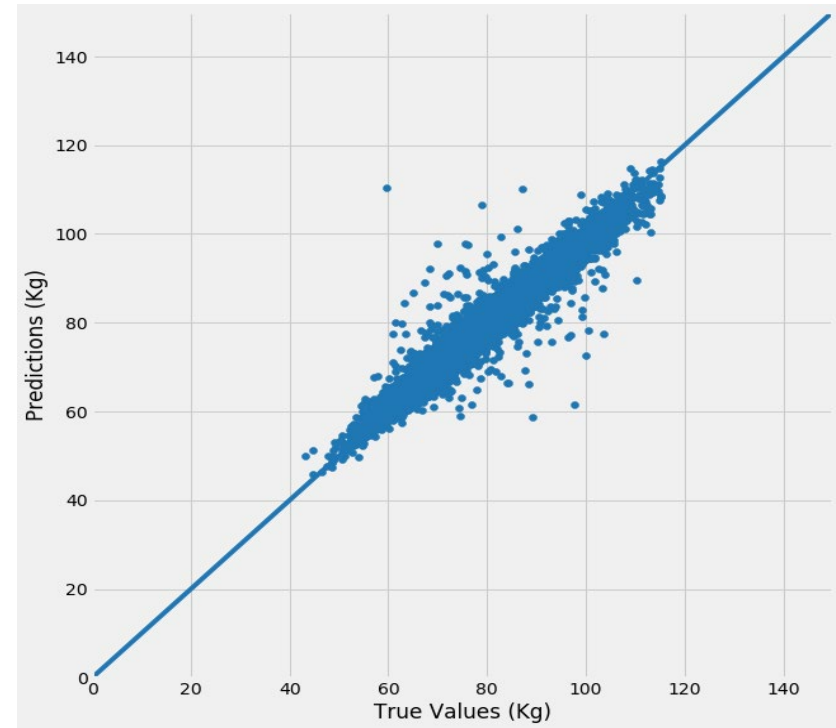
R²: 0.73

1. Phenotypic

vs 2.94 kg

vs 8.4%

0.70



Roasting Cuts

RMSE: 3.07 kg

RRMSE: 3.9%

R²: 0.94

1. Phenotypic

vs 3.23 kg

vs 4.1%

0.93

Summary

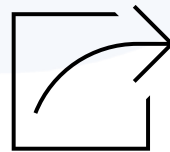
Approach	Data	Algorithm	Grilling		Roasting	
			RMSE	R ²	RMSE	R ²
1	Phenotypic	Gradient Boosting	2.94	0.70	3.23	0.93
2	Carcass Images	Deep Learning	2.84	0.72	3.26	0.93
3	Carcass measurements + Phenotypic	Gradient Boosting	2.78	0.73	3.07	0.94

Conclusions

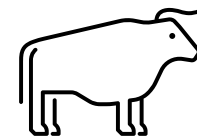
Deep Learning applied to carcass images can predict as well as phenotypic data



Transfer learning is essential for smaller datasets



Domain expertise models outperformed Deep Learning



Our Farmer & Government Representation



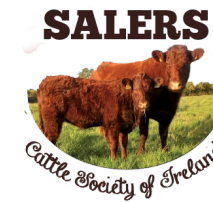
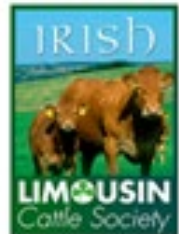
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Bia agus Mara
Department of Agriculture,
Food and the Marine



Our AI & Milk Recording Organisations



Our Herdbooks



MRI Cattle Society of Ireland
Meuse Rhine Issele -- Milk & Muscle!



Norwegian Red Cattle Society



Acknowledging Our Members