# **Breeding livestock for sustainable systems**

Pieter Knap, Katie Olson, Ellen Lai, Matthew Cleveland August 2023





### 'SUSTAINABLE LIVESTOCK SYSTEMS'



Breeding livestock for sustainable systems

Sustainability: classically, the Triple Bottom Line:

People – Planet – Profit

Livestock production: Quadruple Bottom Line:

People – Planet – Profit – PigsPoultryPuminantsPhish



# People – Planet – Profit – PigsPoultryPuminantsPhish **Profit**: selection index: Hazel (1943) food security "feed the globe" "nourish the world" **PigsPoultryPuminantsPhish:** animal welfare



# People – Planet – Profit – PigsPoultryPuminantsPhish

# People:

social justice (e.g. biopiracy: Access & Benefit Sharing) food safety (e.g. cholesterol, PUFA; Salmonella, Listeria etc)

## **Planet**:

resource efficiency environmental efficiency **biodiversity** (e.g. AnGR management)



# People – Planet – Profit – PigsPoultryPuminantsPhish

PigsPoultryPuminantsPhish: animal welfare



Planet: resource efficiency environmental efficiency





- Lotta Rydhmer (16:30 today)
- session 28 (Tuesday afternoon)
- session 77 (Thursday morning)

Environmental efficiency: Greenhouse gas emission





www.recpnet.org/wp-content/themes/recpnet/images/recp/list2.jpg

## livestock production CO<sub>2</sub>eq, worldwide: 15 % of human-made



every livestock sector shows a huge variation in emission intensity (= kg  $CO_2$ eq per kg protein), more so at the higher levels

# How does the carbon footprint of protein-rich foods compare?

beef

lamb

dairy

pork

eggs

fish

Our World in Data Greenhouse gas emissions from protein-rich foods are shown per 100 grams of protein across a global sample of 38,700 commercially viable farms in 119 countries. The height of the curve represents the amount of production globally with that specific footprint. Producing 100 grams of protein from beef The white dot marks the median greenhouse gas emissions for each food product. emits 25 kilograms of CO<sub>2</sub>eq, on average.





#### CO<sub>2</sub>eq per kg protein









#### Good modeling of bad outputs: editors' introduction

Subal C. Kumbhakar<sup>1</sup> · Emir Malikov<sup>2</sup>

...analysis of production technologies when one or more outputs are economically and/or socially undesirable, or so-to-say "bad."

Energies 2020, 13, 5962; doi:10.3390/en13225962



Article **Benchmarking Sustainable Manufacturing:** A DEA-Based Method and Application

Jun-Der Leu, Wen-Hsien Tsai\*, Mei-Niang Fan and Sophia Chuang

Assume there are n decision-making units (DMUs). Each DMU has three factors: inputs, good outputs, and bad (undesirable) outputs, which are represented by three vectors:  $x \in \mathbb{R}^m$ ,  $y^g \in \mathbb{R}^{S1}$ , and  $y^b \in \mathbb{R}^{S2}$ , respectively. The definition of the matrices is from Equations (1)–(9) [46].

$$X = [\chi_1, \ldots, \chi_n] \epsilon$$

$$Y^{g} = \begin{bmatrix} Y_{1}^{g}, \dots, Y_{n}^{g} \end{bmatrix} \epsilon$$
$$Y^{b} = \begin{bmatrix} Y_{1}^{b}, \dots, Y_{n}^{b} \end{bmatrix} \epsilon$$

The possibility set (*P*) can be defined as

$$P = \left\{ \left(\chi, \ y^g, \ y^b\right) \middle| \chi \ge X\lambda, \ y^g \ \le Y^g \ \lambda, \ y^b \ \ge Y^b \ \lambda, \ \lambda \ge 0 \right\}$$
(4)



$\mathbb{R}^{m \times n}$	(1)
$R^{s_1Xn}$	(2)
$R^{s_2Xn}$	(3)

### How to deal with bad outputs

- 1. Reduce the number of bad-output-generating units
- Our *output-generating units* are the animals
- Overhead cost

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- Improved productivity  $\rightarrow$ fewer animals  $\rightarrow$  less bad output
- Improved productivity  $\rightarrow$ more good output
- Milk production, growth rate, leanness, fertility, litter size





How to deal with bad outputs

1. Reduce the number of bad-output-generating units

2. Many non-genetic approaches

3. Make those units more efficient

For example, by improving efficiency by animal breeding



## For example, by improving productivity by animal breeding



### Livestock breeding can influence **direct emissions**:

- Enteric CH<sub>4</sub>
- Manure Management CH<sub>4</sub>
- Manure Management N<sub>2</sub>O





19.2%

Pig breeding can influence **direct emissions**:

- Enteric CH<sub>4</sub>
- Manure Management CH<sub>4</sub>
- Manure Management N<sub>2</sub>O





3.1% /

**`27.1%** 

12.7%

Animals vary in terms of their amino acid requirements, due to variation in

- gross body protein deposition (~ lean tissue growth rate)
- body protein composition  $\bullet$ 
  - muscle
  - connective tissue
  - gastro-intestinal
  - other tissues
- net efficiency of nitrogen metabolism



Figure 7 - Estimation of individual daily lysine requirements (in g/mJ net energy [NE]) in growing-finishing pigs.



Pomar et al (2009) Rev Bras Zootec 38 suppl



Genus

pigs

The Premier Journal and Leading Source of New Knowledge and Perspective in Animal Science

Nitrogen excretion at different stages of growth and its association with production traits in growing pigs M. Shirali, A. Doeschl-Wilson, P.W. Knap, C. Duthie, E. Kanis, J.A.M. van Arendonk and R. Roehe

J ANIM SCI 2012, 90:1756-1765.

#### Nitrogen Utilization Efficiency

- heritability: 0.31 to 0.41
- r<sub>G</sub> with FCR: around –0.95

Van der Peet (1999); Shirali (2012); Saintilan (2013); Ali (2019); Soleimani (2020); Kasper (2021).

# FCR is a regular selection trait in pig breeding

Proxy !





Nitrogen Utilization Efficiency is heritable and strongly correlated to FCR On the farm level, whole-enterprise FCR (FCR<sub>we</sub>) is determined by

- growth rate, feed intake & pig mortality of the market pigs
- feed intake, litter size & piglet mortality of their mothers •

Regular selection traits in pig breeding

... included in the routine selection index based on their Marginal Economic Values

... MEVs, based on maximization of farmer profitability

... not on minimization of the footprint

...but that could be changed: footprint index



#### Contribution of traits to the variation in routine and footprint indexes





### **Correlations?**





50

30

10

-10

(kg/pig)

Ъ

CO2e







![](_page_19_Picture_1.jpeg)

![](_page_20_Figure_0.jpeg)

trait	line	slope	11 years
CO <sub>2</sub> eq -	D1	-1.25	
	D2	-3.20	
	S1	-4.28	
	S2	-2.97	

Selection on the routine index reduced the lifetime footprint of the slaughter pig by 32.2 kg CO<sub>2</sub>eq in 11 years: 2.93 kg / year.
Total lifetime footprint: ~ 300 kg CO<sub>2</sub>eq.
We reduce the footprint by about 1 % per year, by selecting on index scenarios that were never designed to tackle the footprint.

![](_page_20_Picture_3.jpeg)

![](_page_21_Figure_0.jpeg)

![](_page_21_Picture_1.jpeg)

Capper et al (2009, 2019) bovidiva.com Greenhouse gas emissions (CO<sub>2</sub>eq)

- Scope 1 & 2: internal
- Scope 3: external, upstream supply chain

![](_page_22_Picture_3.jpeg)

![](_page_22_Picture_4.jpeg)

![](_page_22_Picture_5.jpeg)

![](_page_22_Picture_6.jpeg)

Greenhouse gas emissions (CO<sub>2</sub>eq)

- Scope 1 & 2: internal ullet
- Scope 3: external, upstream supply chain ullet

![](_page_23_Figure_3.jpeg)

![](_page_23_Picture_4.jpeg)

#### 150 food-processing companies

Greenhouse gas emissions ( $CO_2eq$ )

Scope 1 & 2: internal •

ABS

Genus

Scope 3: external, upstream supply chain  $\bullet$ 

![](_page_24_Figure_3.jpeg)

### Genus 2022: CO<sub>2</sub>

17 % (Genus 2022)

scope 3

0.1

0

-0.1

scope 1,2

scope 3:

Mton

CO<sub>2</sub>eq

Greenhouse gas emissions (CO<sub>2</sub>eq)

- Scope 1 & 2: internal
- Scope 3: external, upstream supply chain
- Scope 3: external, downstream

Livestock breeding (e.g. Genus 2022):

- reduction of scope 3 downstream CO<sub>2</sub>eq
   = 12 (16 ?) × absolute internal & upstream CO<sub>2</sub>eq
- disregarding the emission by multiplier farms
- downstream is about  $\Delta G$ , not about physical animals

-0.7 scope 1,2,3: **absolute** emissions, internal & upstream

scope 3

0.1

0

-0.1

-0.2 -

-0.3 -

-0.4

-0.5 ·

-0.6

scope 1,2

Mton

CO<sub>2</sub>eq

![](_page_25_Picture_9.jpeg)

### scope 3: **reduction** of global downstream emissions due to routine genetic improvement

![](_page_25_Picture_12.jpeg)

### Genus 2022: $CO_2$ and $\Delta CO_2$

Farm-level footprint, based on FCR<sub>we</sub>

- Reduction of the footprint through regular selection: 1% per year
- 24 years to achieve 24 % mitigation
- Question: how long will it take to shift all pig producers to the top-10 % level, for any non-genetic criterion?
- Exactly...
- Conclusion: genetic improvement has an important role to play

![](_page_26_Figure_6.jpeg)

- Mitigation potential: if all producers would apply the practices of the 10 % producers with the lowest emission intensity (no output reduction) • 37 % mitigation for cattle, 24 % for pigs
- FAO always ignores genetic improvement
- so those practices are non-genetic

![](_page_26_Picture_12.jpeg)

![](_page_26_Figure_14.jpeg)

We reduce the footprint by 1 % per year, by selecting on routine indexes that were never designed to tackle the footprint.

Question 1: is that enough? Answer: depends who you ask.

Sooner or later, someone will argue that it is **not** enough. How do we deal with that?

Options:

- 1. Arbitrarily increase focus on growth rate and feed intake in the routine index scenarios
- 2. Include the farmer's cost of GHG emission into the trait MEVs Shadow price of carbon
- The price of a license to emit a ton of CO<sub>2</sub> into the atmosphere
- The tax levied on the emission of a ton of  $CO_2$  into the atmosphere If the farmer has a financial incentive for reducing his footprint, then we can work that into his profit equation  $\rightarrow$  into the trait MEVs

![](_page_27_Picture_8.jpeg)

![](_page_27_Figure_9.jpeg)

![](_page_28_Figure_0.jpeg)

![](_page_28_Picture_1.jpeg)

carbonpricingdashboard.worldbank.org/map\_data en.wikipedia.org/wiki/Carbon\_emission\_trading

![](_page_28_Figure_3.jpeg)

Genu

# Effects of incorporating environmental cost and risk aversion on economic values of pig breeding goal traits

B.M. Ali<sup>1</sup> | Y. de Mey<sup>1</sup> | J.W.M. Bastiaansen<sup>2</sup> | A.G.J.M. Oude Lansink<sup>1</sup>

### A simulated breeding program with 1 round of selection

		MEV		
trait	0	€ 38	increase	trait
growth rate	0.0649	0.0701	8.0%	GHO
FCR	17.1485	19.0219	10.9%	(kg C
litter size	1.9743	2.0645	4.6%	
piglet mortality	0.2820	0.2964	5.1%	
PIC ABS		$\sim$ shadow price of carbon (f / tCO_eq)		

	hoforo	af	
trait	belore	0	
GHG emission	221	_1 1	
(kg CO2eq / pig)	221	-1.1	
	reductio	on : 1.1	

![](_page_29_Picture_7.jpeg)

![](_page_29_Picture_8.jpeg)

#### a shadow price of € 38 is not effective

![](_page_29_Figure_10.jpeg)

- .131 / 221 = 0.51 % / year
- 24 % in 47 years
- 24 % in 24 years

Energy Policy 109 (2017) 288–296

![](_page_30_Picture_1.jpeg)

Contents lists available at ScienceDirect

**Energy Policy** 

journal homepage: www.elsevier.com/locate/enpol

### Worldwide carbon shadow prices during 1990–2011

Jean-Philippe Boussemart<sup>a</sup>, Hervé Leleu<sup>b</sup>, Zhiyang Shen<sup>c,\*</sup>

Macro-economic approach, 119 countries:  $\sim$  regress each country's annual GDP (\$) on its annual CO<sub>2</sub> emission volume

![](_page_30_Picture_8.jpeg)

![](_page_30_Picture_9.jpeg)

![](_page_31_Figure_0.jpeg)

# $\Delta$ (GDP) with a lower $\Delta$ (GHG)

Boussemart et al. (2017)

- include the cost of CO<sub>2</sub> mitigation into the profit equation to calculate the trait MEVs  $\bullet$
- shadow price of carbon
- current shadow prices are defined politically
- seriously lower than the true macroeconomic values – understandably
- current levels are far too low to create a  $\bullet$ realistic incentive for animal breeding

![](_page_32_Picture_5.jpeg)

![](_page_32_Picture_6.jpeg)

![](_page_32_Figure_7.jpeg)

![](_page_32_Figure_9.jpeg)

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![](_page_33_Picture_2.jpeg)

![](_page_33_Picture_3.jpeg)

### 'SUSTAINABLE LIVESTOCK SYSTEMS'

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