

# Variable selection

# Does it matter from a machine learning perspective?

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#### **Traditional statistics**



Two groups of pigs: diet A, diet B  $\rightarrow$  which promotes growth better?

weight = mean + diet + e

!! is "diet" relevant/significant for growth?

Diet A	Diet B
90 kg	89 kg
88 kg	82 kg
92 kg	79 kg
87.5 kg	83 kg

#### **Traditional statistics**



What about the sex of the pigs? And their age? Or breed? Interactions?

weight = mean + diet + sex	+ age + e
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weight = mean	+ diet + sex + age	+ breed + + e
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weight = mean + diet + sex	+ age + breed +	age*breed + + e
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Diet A	Diet B
90 kg	89 kg
88 kg	82 kg
92 kg	79 kg
87.5 kg	83 kg

# What about machine learning?



In machine learning the model learns on its own which variables to use and how (not easily accessible by humans)

"black box"

# What about machine learning?



In machine learning the model learns on its own which variables to use and how (not easily accessible by humans)



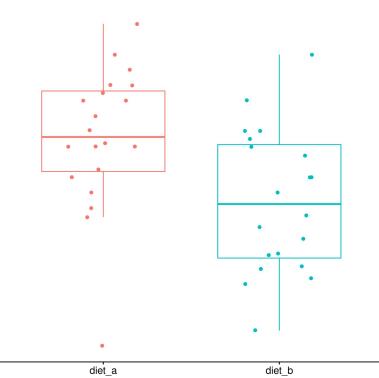
- If the ML model is able to determine that burying ground quartz stuffed into the horn of a cow (which are said to harvest "cosmic forces in the soil") has no effect on the growth of pigs, this information is kept hidden in the model
- Sounds suboptimal, but the interpretability of single variables loses sense as the number of variables and their combinations and transformations increases
- And in modern statistics we usually have A LOT of variables!

#### **An illustration - ANOVA**



*term	SS	d.f.	F	p-value
intercept	189.2	1	415.11	5.2e-37
diet	2.35	1	5.39	0.022

100



variable 🖮 diet\_a Ė diet\_b

\*made up numbers!!

4.79

residual

variable

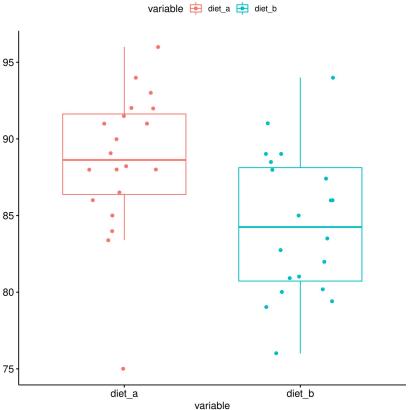
75 -



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=

weight = mean + b1\*diet + e





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```
*mean = 80 kg
*b1 = +2.75 kg [coding diet A = 1; diet B = 0]
```

#### **Interpretation**

- mean: average weight of pigs
- b1: average difference in weight between pigs fed with diet A and pigs fed with diet B

\*made up numbers!!





weight = mean + b1\*diet + b2\*age + e

```
mean = 80 kg
b1 = +2.75 kg [coding diet A = 1; diet B = 0]
p-value = 0.006 \rightarrow b2 = +1.47 kg [coding age in years]
```

#### **Interpretation**

- mean: average weight of pigs
- b1: average difference in weight between pigs fed with diet A and pigs fed with diet B
- b2: average weight gain per year of age, keeping diet constant

made up numbers!!



weight = mean + b1\*diet + b2\*age + b3\*motion\_time + e

Traditional statistics approach!

The coefficient for "motion time" (b3) has a p-value of 0.29: we decide not to include motion time in the model

But what if the relationship between motion time and weight is not linear? We can fit **polynomial terms**! (square, cube etc.) [  $\rightarrow$  this is still a linear model!]



weight = mean + b1\*diet + b2\*age + b3\*motion\_time + b4\*motion\_time<sup>2</sup> + b5\*motion\_time<sup>3</sup> + e

The p-values for the polynomial terms are now 0.075, 0.051 and 0.032:

- should we include these in the model?

The coefficients for the polynomial terms are: -1.57, 0.24 and -0.03

- how should we interpret these?
- on average, we lose 1.57 kg per hour of motion, we gain 0.24 kg per hour-of-motion squared, and we lose 30 grams per hour-of-motion cubed

How to build and interpret the model becomes more and more confused

## The machine learning perspective



With many variables (but already with a handful of variables) it becomes a titanic task to decide which variables, combinations of variables and functions of variables include in the model

→ let the model decide!

The questions of variable selection and model interpretability become ill-posed

→ predictions matter more than inference!

Is this the end of the story? Can we really say nothing about why our model works (or does not work)?

→ don't panic, we'll be able to crack the black box (at least partially)

# The machine learning perspective



variable selection ≠ data representation

variable selection ≠ feature engineering

How does the model decide which variables to use? (hold on a little longer ...)