Data management in precision farming

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Adequation between question and data

- What can I do with existing data? (meta analysis)
- What's the question ?
  - $\Rightarrow$ Will my data answer it (write the method+stats first)

Adequation between question and data

- Practical decision
  - Real time (Importantn for large farms)
    - On farm alerts (vêlage, chaleurs, . . .)
  - Long term
    - Herd management tools
    - Planning asistants
- V.S. Research question
  - Impact of feeding on production/environment
  - Gentic selection on food efficiency?
  - Animal behaviour/welfare?

Costs v.s. benefits

Paris 1889, la pose du troisième étage de la Tour de Monsieur Eiffel.
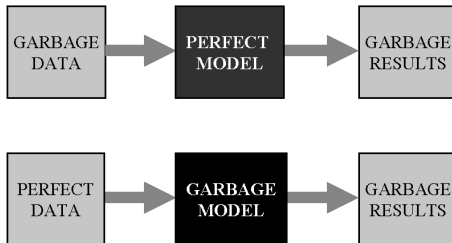
Plonk et Replonk

⇒ Cost v.s. benefits tradeoff
- money
- time
- human factors

Get consistent data

MODEL CALCULATIONS
"Garbage In-garbage Out" Paradigm

- Farm proof systems (machine robustness)
- Know what you measure
  - What's the data (definition, format, unit)
  - In which context it's measured / usable ?

Organize data

- Use databases (SQL is your friend)
    - SQL https://www.w3schools.com/sql/
    - SQLite in R http://cpc.cx/iZj
- Maximize automation (program boring repetitive stuff)
    - With R http://cpc.cx/kQJ
    - With python http://cpc.cx/kQK
    - Others softwares are OK
- Write documentation
    - Someday, it will save someone
    - Who may be you.

Watch your data

- Make plots
  - R `http://cpc.cx/kQL`
  - R+ggplot2 `http://cpc.cx/kQM`
  - R+shiny `https://shiny.rstudio.com/`
- Descriptive stats
  - mean, median, in, max. . .
  - proportion of missing values
  - R summary
- PCA. . .

Principal Component Analysis

minimal variance

maximal variance

Sadly not my little sister

- PCA maximize the projected variance
  - ▸ Orhtogonal axis
  - ▸ R, factominer http://cpc.cx/kQQ
- PCA is a good tool for
  - ▸ Summarizing data (check inertia)
  - ▸ Finding outliers (check individual weight)
  - ▸ Constructing synthetic variables (check axis contribution)
- PCA is NOT a tool for
  - ▸ Acurate correlations ⇒ watch the correlation matrix
  - ▸ Testing hypothesis ⇒ statistical test

Data preprocessing

http://www.tylervigen.com/spurious-correlations
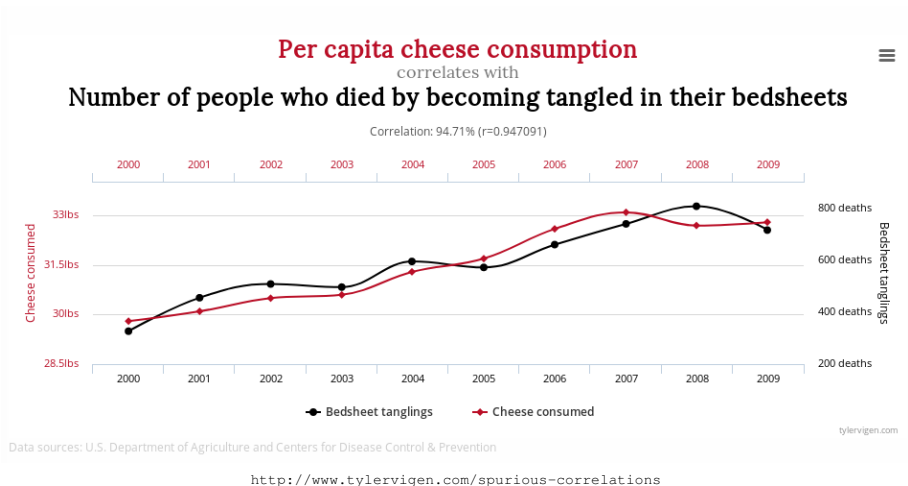
Data preprocessing

Calvin and Hobbs, Bill Watterson

- Remove unrelevant variables
    - Remove unrelated variables (spurious correlation)
        - Crucial on datasets with many variables
    - Remove a posteriori variables
   - crash survival day7 = alcool+ crash speed + ~~reeducation duration~~
- Remove unrelevant data
    - Out of scope data (ex : holstein model, montbeliarde data)
    - Outliers (PCA, boxplots, . . .)
        - Wrong data (ex broken sensor) ⇒ discard
        - Rare events (ex mastitis) ⇒ kept or not?

Data preprocessing

- Data smoothing
- OK when we expect few changes within neighbors (time, space)
- Methods
    - kernel methods (ex : rowing average)
    - rowing median
    - splines
    - . . .

Data preprocessing
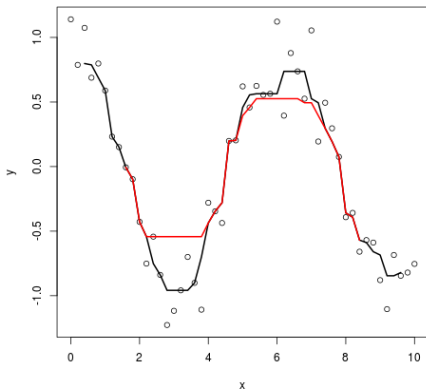
median filter, black 5 points, red 17 points

- Smoothing strength is a compromise.
- That depends on your question/model.
- An on your data variability.

milky way, with increasing size gaussian kernel

milky way, with increasing size gaussian kernel

milky way, with increasing size gaussian kernel

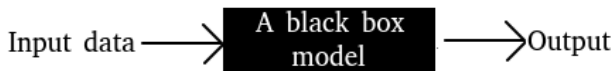milky way, with increasing size gaussian kernel

Aggregate variables

When?

- The same thing is measured multiple times
  - Ex : captor1 and captor2
  - Rare gold standard v.s. frequent low quality measurments
- Strong correlation between variables
  - Ex behavioural traits

How?

- Choose a variable
- Do a variable "mix" (ex linear combination)
- PCA : synthetic variables (a linear combination)
  - Don't overinterpret PCA synthetic variables
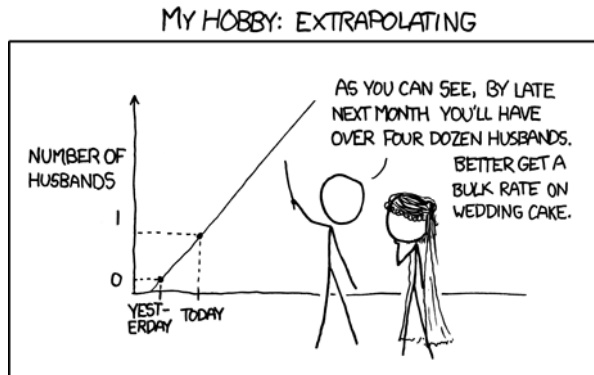
Is my model OK or garbage?

Input data ⟶ A black box model ⟶ Output

- A model can be anything
  - Linear model http://cpc.cx/kQE , http://cpc.cx/kQF
  - Non linear model http://cpc.cx/kQG
  - A classifier http://cpc.cx/kQR
  - . . .
  - The devil himself in a unicorn disguise
- ⇒ Garbage in, garbage out
- We assume that input data is ok, is the model garbage?

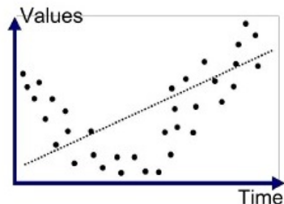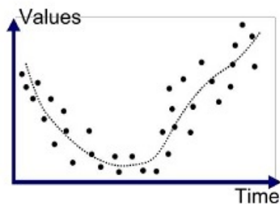# Typical evil models : extrapolation

- Your training and testing data must cover your practical use cases

Typical evil models : Overfitting
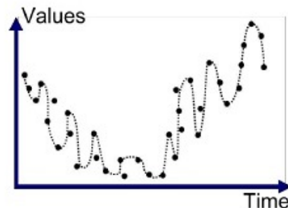

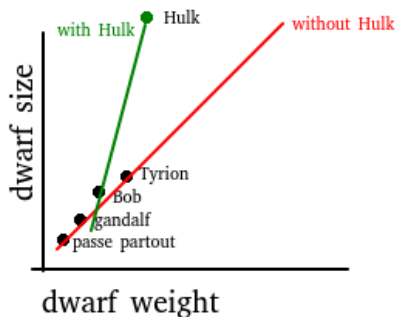Underfitted   Good Fit/Robust   Overfitted

- Many parameters, few data ⇒ Check degree of freedom

Typical evil models : outliers

- Outlier with heavy weight ⇒ Check for outliers
- Check your question
  - ▶ General model : keep Hulk, rename axis
  - ▶ Dwarf model : discard Hulk

WARNING : evil models exists in unexpected cases
⇒ ALWAYS TEST YOUR MODELS

Test your models

	detect mine	detect rock			detect mine	detect rock
mine	9	1		mine	5	5
rock	50	50		rock	10	90

Two mines detectors, what's the best one?

- All mistakes are NOT equivalent
  - ▶ Be carefull, some methods(like area under ROC curve) say so
  - ▶ When possible use a real-life criterion (human lives, money, ...)
  - ▶ When not, define a a reasonable criterion **apriori**.

Test your models

- What the difference between model prediction and model output
    - Plot observed v.s. predicted
    - Model fitting (ex sum of square)
    - Cross with other knowledge source (common sense, litterature, meta analysis ...)
- Model testing
    - Use an **independant** dataset
    - Test the model on it, according to your criterion
    - See bootstrap like methods.

Conclusion

Practical or scientific question

Which data should be collected ?

SQL, doc, programs → Aggregate data

Plots, stats PCA → Visualize data — *Is everythin OK? New ideas?*

PCA correlations → Filter variables — *Keep variables relevant to the question*

PCA, Boxplots, Smoothing → Preprocess data — *Remove outliers, Smooth*

Lm, glm, others models → Write and fit a model — *Evil models (extrapolation, Over-fitted, outlier)*

Observed vs predicted, independent dataset → Test a model — *Clever "good enough" criterion*