

# Relative importance of steps

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# Relative order of importance

question > data > features > algorithms

## An important point

The combination of some data and an aching desire for an answer does not ensure that a reasonable answer can be extracted from a given body of data.

John Tukey

# Garbage in = Garbage out

question -> input data -> features -> algorithm -> parameters -> evaluation

1. May be easy (movie ratings -> new movie ratings)
2. May be harder (gene expression data -> disease)
3. Depends on what is a “good prediction”.
4. Often more data > better models
5. The most important step!

# Features matter!

question -> input data -> features -> algorithm -> parameters -> evaluation

## Properties of good features

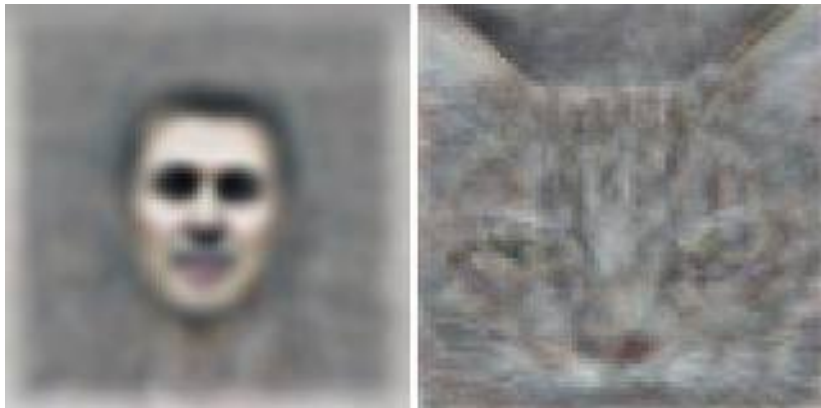
- ▶ Lead to data compression
- ▶ Retain relevant information
- ▶ Are created based on expert application knowledge

## Common mistakes

- ▶ Trying to automate feature selection
- ▶ Not paying attention to data-specific quirks
- ▶ Throwing away information unnecessarily

## May be automated with care

question -> input data -> features -> algorithm -> parameters -> evaluation



<http://arxiv.org/pdf/1112.6209v5.pdf>

# Algorithms matter less than you'd think

question -> input data -> features -> algorithm -> parameters -> evaluation

TABLE 1

*Performance of linear discriminant analysis and the best result we found on ten randomly selected data sets*

Data set	Best method e.r.	Lindisc e.r.	Default rule	Prop linear
Segmentation	0.0140	0.083	0.760	0.907
Pima	0.1979	0.221	0.350	0.848
House-votes16	0.0270	0.046	0.386	0.948
Vehicle	0.1450	0.216	0.750	0.883
Satimage	0.0850	0.160	0.758	0.889
Heart Cleveland	0.1410	0.141	0.560	1.000
Splice	0.0330	0.057	0.475	0.945
Waveform21	0.0035	0.004	0.667	0.999
Led7	0.2650	0.265	0.900	1.000
Breast Wisconsin	0.0260	0.038	0.345	0.963

<http://arxiv.org/pdf/math/0606441.pdf>

## Issues to consider

### The “Best” Machine Learning Method

**Interpretable**

**Simple**

**Accurate**

**Fast**  
(to train and test)

**Scalable**

[http://strata.oreilly.com/2013/09/  
gaining-access-to-the-best-machine-learning-methods.  
html](http://strata.oreilly.com/2013/09/gaining-access-to-the-best-machine-learning-methods.html)



# Prediction is about accuracy tradeoffs

- ▶ Interpretability versus accuracy
- ▶ Speed versus accuracy
- ▶ Simplicity versus accuracy
- ▶ Scalability versus accuracy

# Interpretability matters

```
if total cholesterol  $\geq 160$  and smoke then 10 year CHD risk  $\geq 5\%$   
else if smoke and systolic blood pressure  $\geq 140$  then 10 year CHD risk  $\geq 5\%$   
else 10 year CHD risk  $< 5\%$ 
```

<http://www.cs.cornell.edu/~chenhao/pub/mldg-0815.pdf>

# Scalability matters



Innovation

by [Mike Masnick](#)

Fri, Apr 13th 2012  
12:07am

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## Why Netflix Never Implemented The Algorithm That Won The Netflix \$1 Million Challenge

from the *times-change* dept

You probably recall all the excitement that went around when a group **finally won** the big Netflix \$1 million prize in 2009, improving Netflix's recommendation algorithm by 10%. But what you might *not* know, is that **Netflix never implemented that solution itself**. Netflix recently put up a blog post **discussing some of the details of its recommendation system**, which (as an aside) explains why the winning entry never was used. First, they note that they *did* make use of an earlier bit of code that came out of the contest:

<http://www.techdirt.com/blog/innovation/articles/20120409/03412518422/>

<http://techblog.netflix.com/2012/04/netflix-recommendations-beyond-5-stars.html>