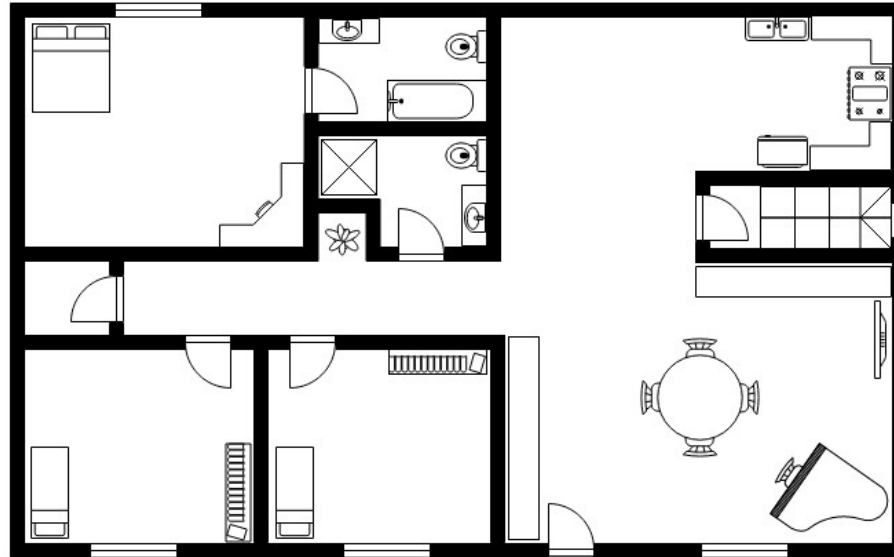


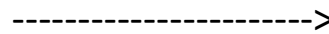
# Machine Learning

- Learning from Data
- Models
- Features and Targets
- Dimensions of Machine Learning

## Estimating Apartment Prices



Features?



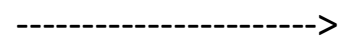
Target: Price

Regression: Predicting a real number value

## Kind of Iris Plant



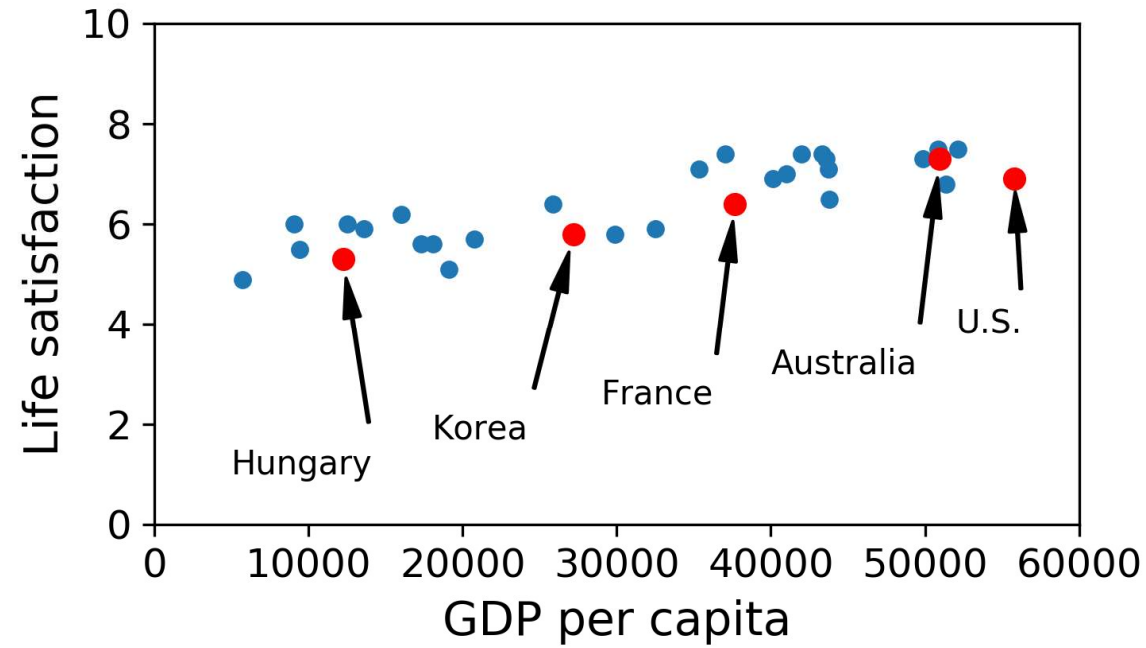
Features?



Target: Kind

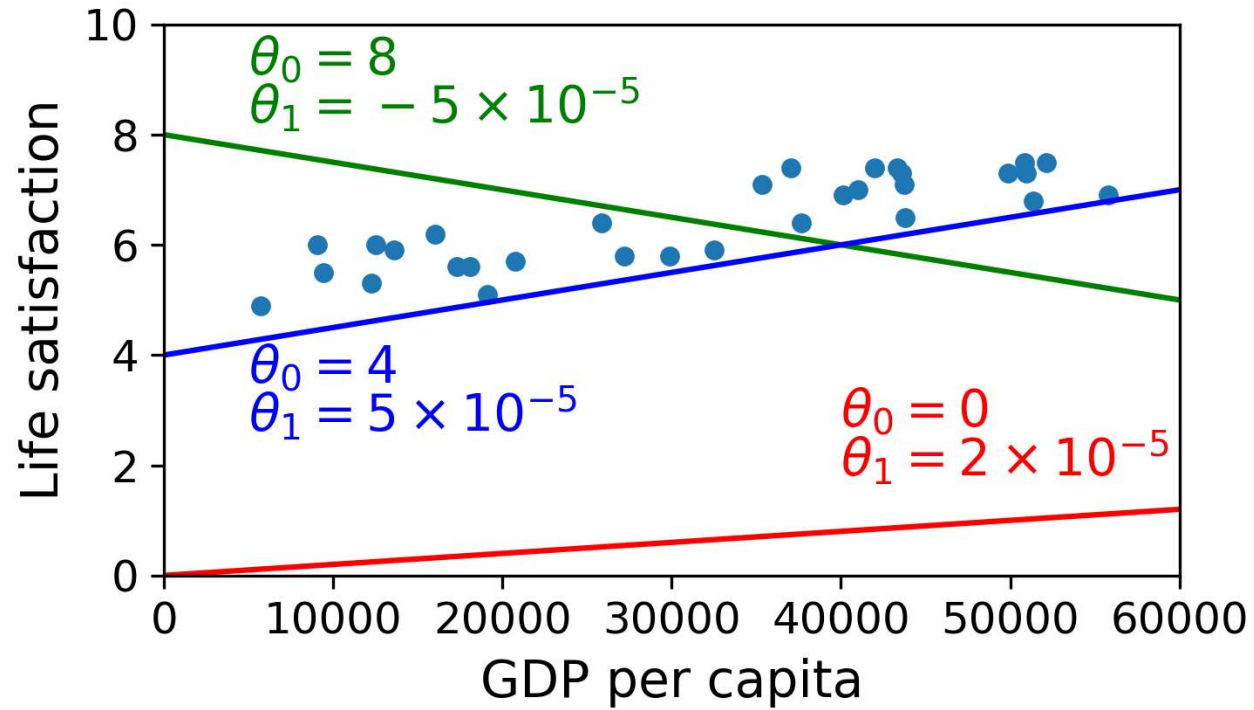
Classification: Predicting a label

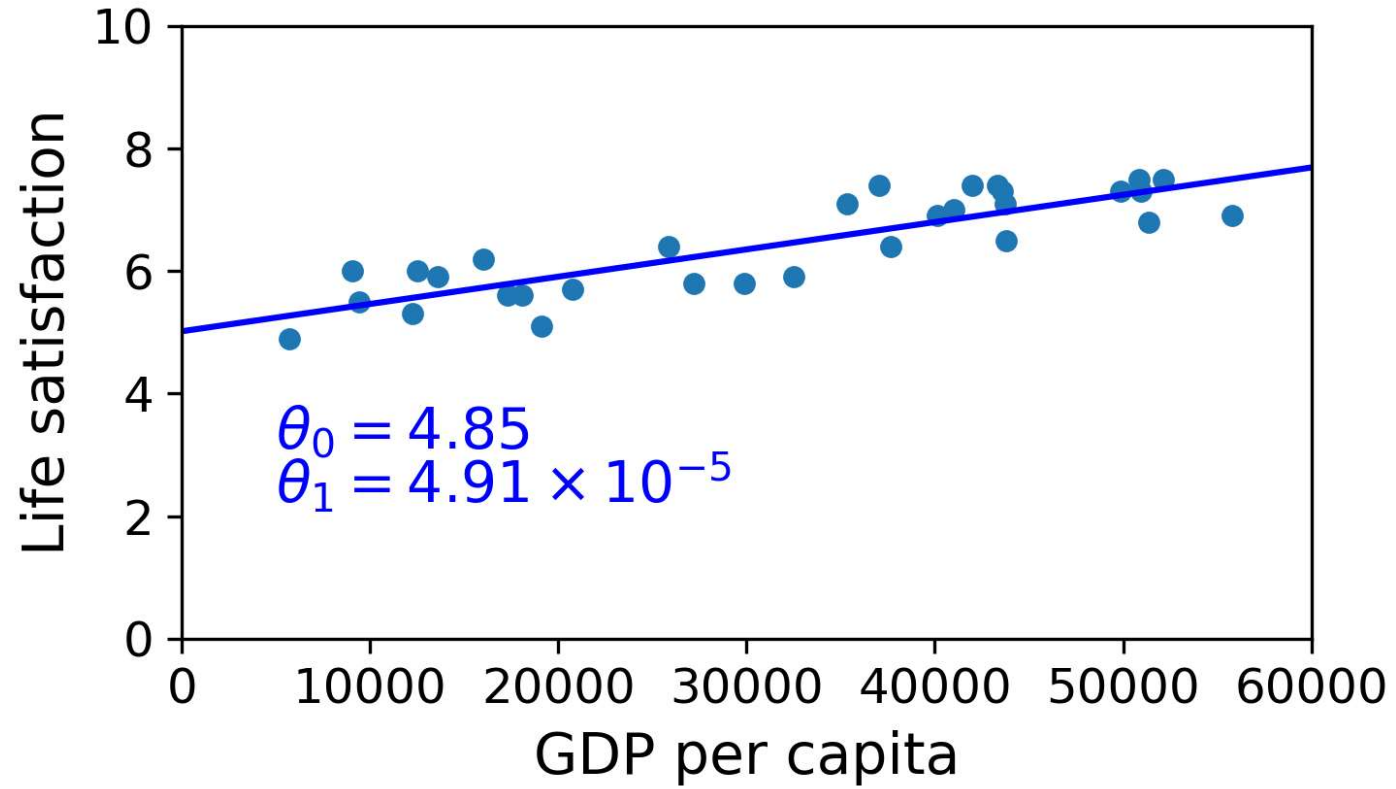
Is there a trend?

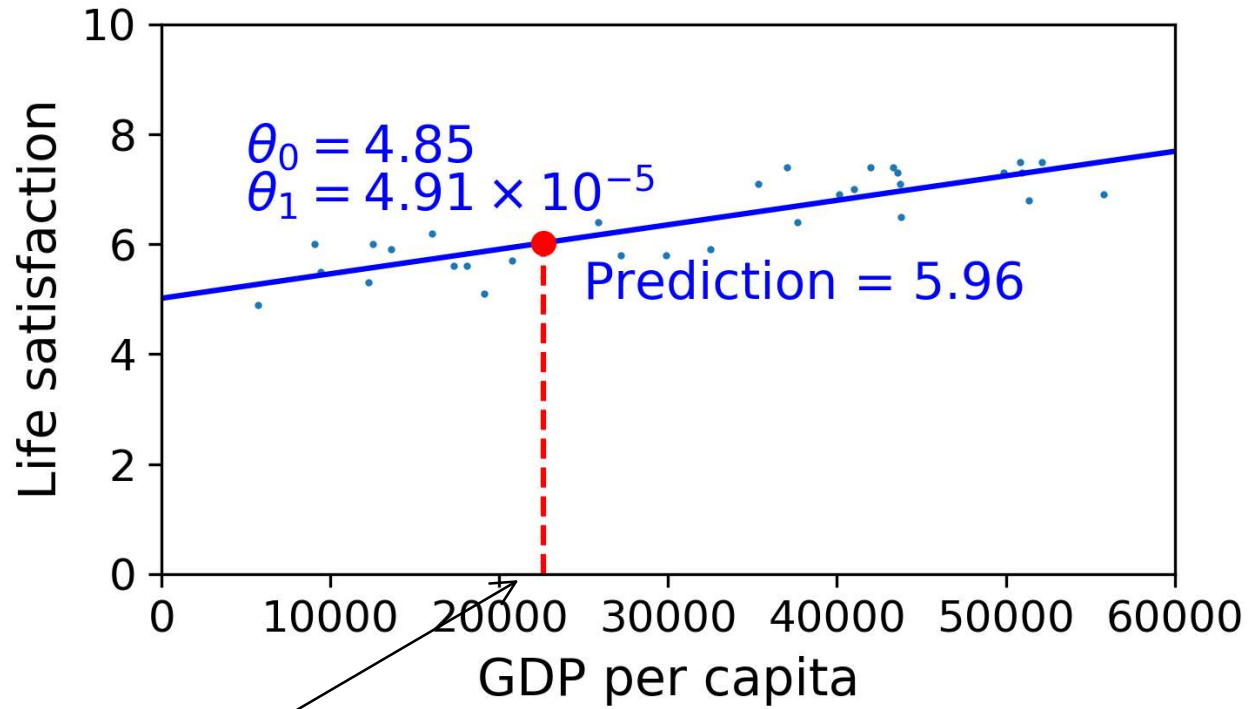


Linear model:  $sat = \theta_0 + \theta_1 * gdp\_per\_capita$

Finding good models?  
How is "good" defined





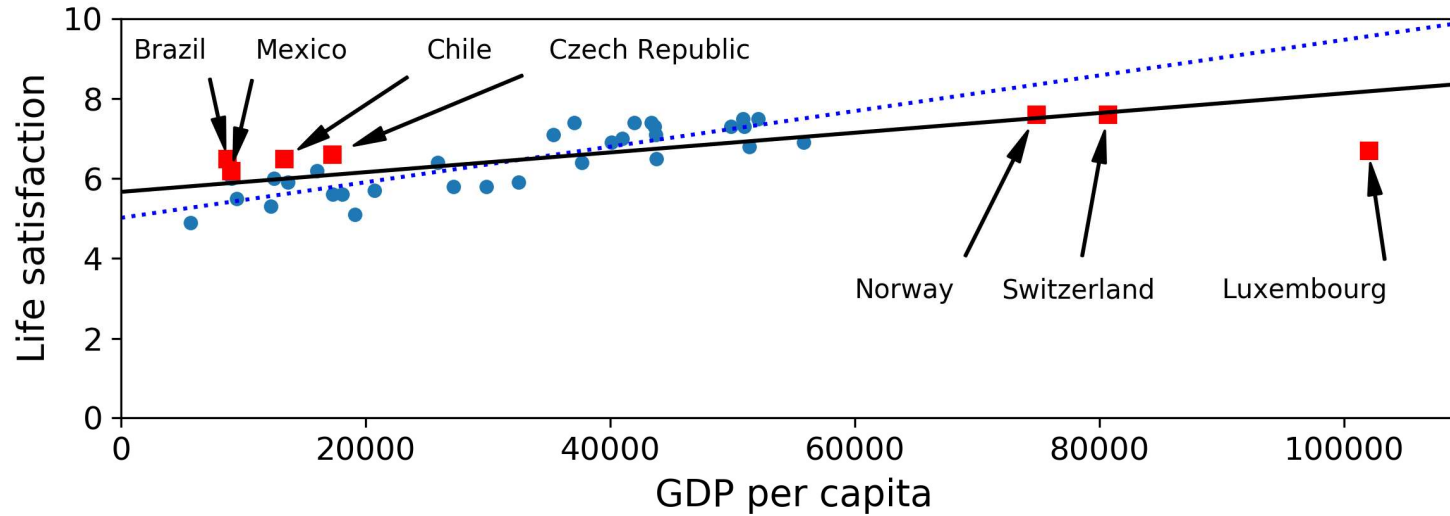


Take GDB per capita and predict sat

- Not enough data
- Data is not representative
- Bad data quality
- Irrelevant features
- Overfitting
- Underfitting

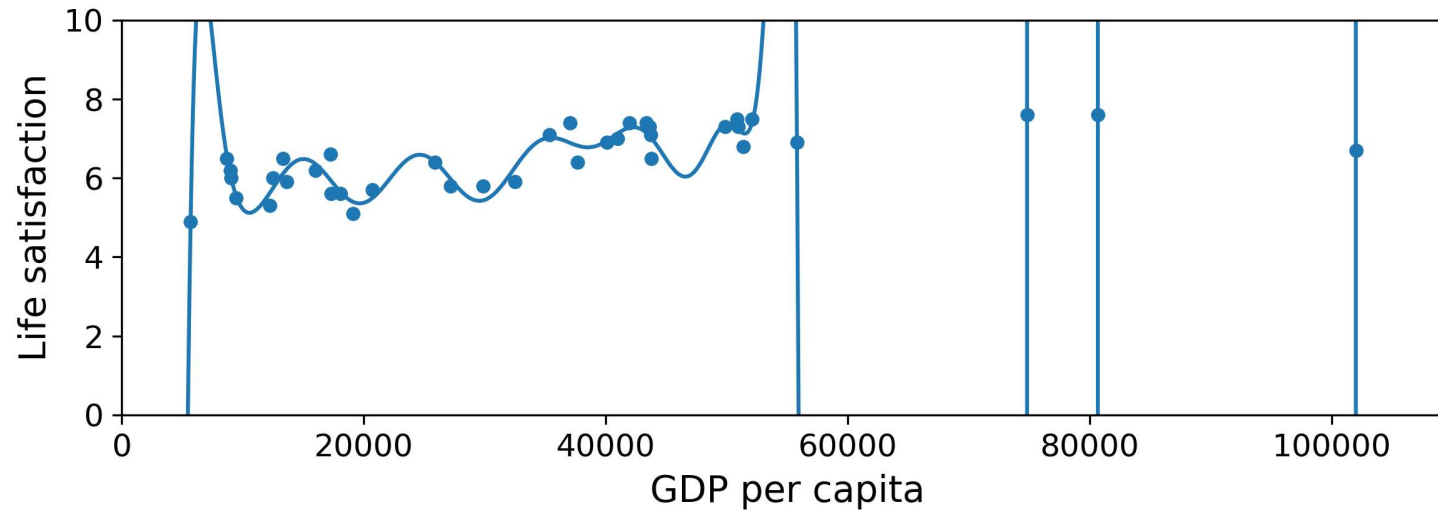


Adding Brasil, Mexico, ..., Norway, ... changes model  
 Linear model does not really fit



Countries with high GDP per capita but lower sat  
 Countries with low GDP per capita but higher sat

Komplex model  
Polynomial with high degree



## Goal:

- Find relationship between features and targets
- Discover patterns
- Find correlations

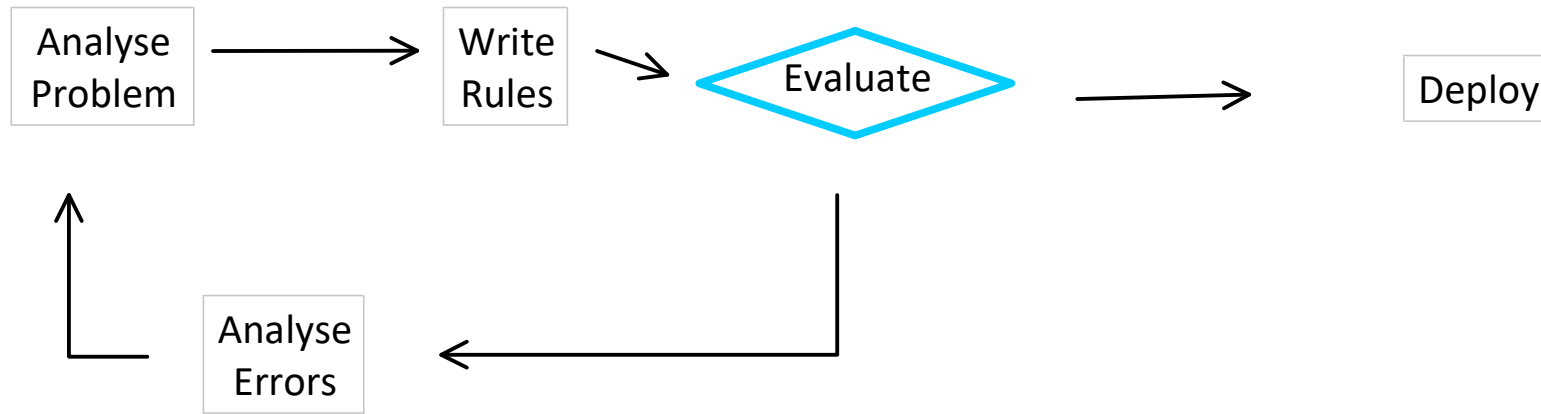
## Common terms:

- Machine Learning
- Data Mining
- Statistical Learning

## Reasons why learning from data

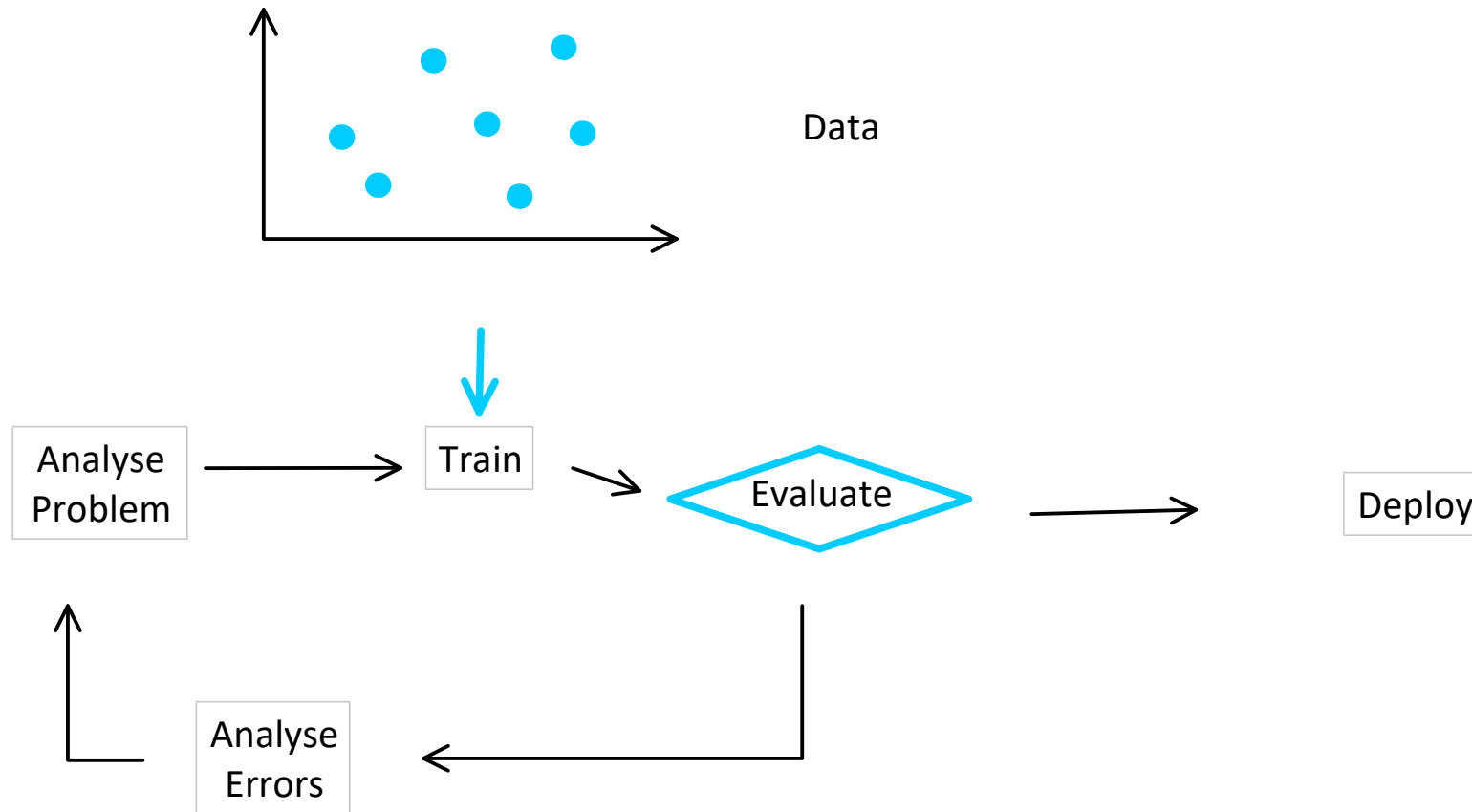
- Constructing an abstraction for the relationship between features and targets (model)
- Complex problems where algorithmic formulation is hard
- Environments where data changes continuously
- Problems that require a long list of complicated rules

Analysis of spam emails, identification of offensive words and phrases



Long list of rules - not easily maintainable  
Adaption to changes are hard to do

Identification auf words and phrases by learning from data using suitable algorithms.



Adaption to changes much easier: just take new data and learn again

## Features

- Attributes that characterize a particular instance/object
- Also called predictors

## Targets

- Scalar value in case of regression
- Nominal value in case of classification

## Features / Targets

### Regression

Bedrooms	Bathrooms	Latitude	Longitude	Price
1	1	40.71	-73.94	3055
...	...	...	...	...

### Classification

Income	Job	IsMarried	Age	Loan
...	...	...	...	Yes
...	...	...	...	No

<b>x1</b>	=	x11	x12	x13	...	x1n	y1
<b>x2</b>	=	x21	x22	x23	...	x2n	y2
...		...					...
<b>xm</b>	=	xm1	xm2	xm3	...	xmn	ym

**X**

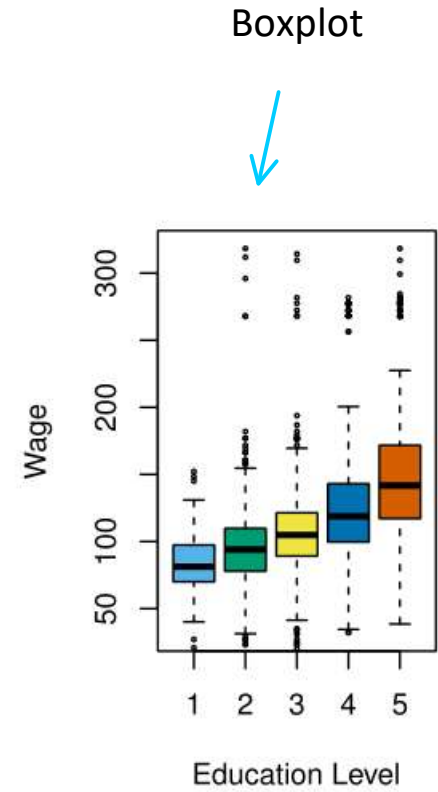
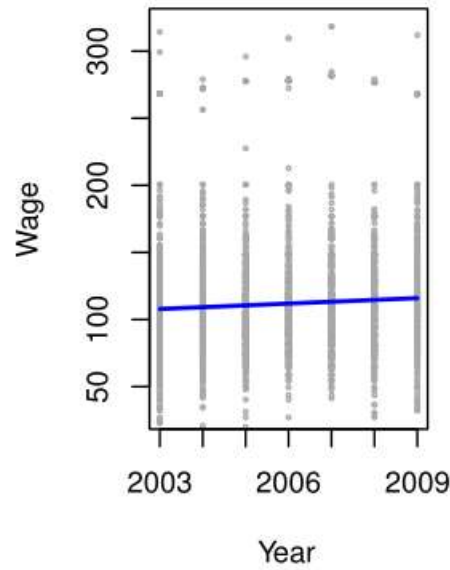
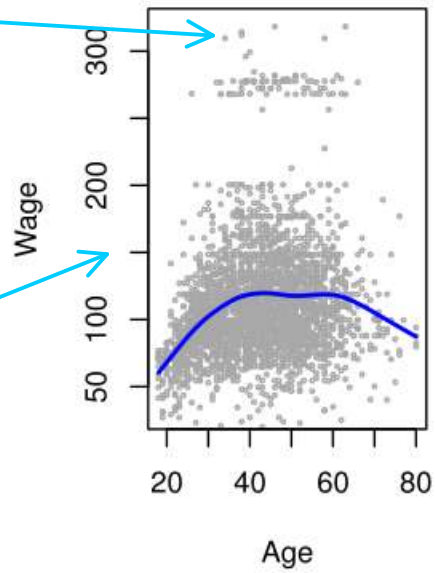
**y**

<b>x<sub>i</sub></b>	Feature vector
<b>x<sub>ij</sub></b>	Feature value
<b>y<sub>i</sub></b>	Target value
<b>X</b>	Feature matrix
<b>y</b>	Target vector

- All entries must be numbers, i.e. all feature and target values must be transformed into numbers
- Each feature vector is one instance/object

each point is  
one person  
(object, instance)

Scatterplot





Learning should

- be as accurate as possible
- generalize to new data as good as possible

Conflicting goals!

Just remember target value wrt. feature combination

- Good accuracy
- Bad generalization
- Combinations might not exist
- Combinations might not be unique

bedrooms	bathrooms	latitude	longitude	price
0	1.0000	40.7073	-73.9664	2650
0	1.0000	40.7073	-73.9664	2850
0	1.0000	40.7073	-73.9664	2950
0	1.0000	40.7073	-73.9664	2850

Build coarse grained groups, e.g. price per bedrooms and bathroom combinations

- Bad accuracy
- Good generalization

Bedrooms	Bathroom	Price
1	1	3000
2	1	3700
1	2	3010
...	...	...

Learn from neighbors (k nearest neighbor)

- Powerful
- But must store complete training data set
- No real learning
- Finding nearest neighbors requires search, might be slow

Build a linear model:

$$\text{Price} = w_0 + w_1 * \text{bedrooms} + w_2 * \text{bathrooms} + w_3 * \text{latitude} + w_4 * \text{longitude}$$

Might be not suitable: higher number of bathrooms doesn't mean higher price

## Supervision

- Supervised
- Unsupervised
- Semisupervised
- Reinforcement Learning

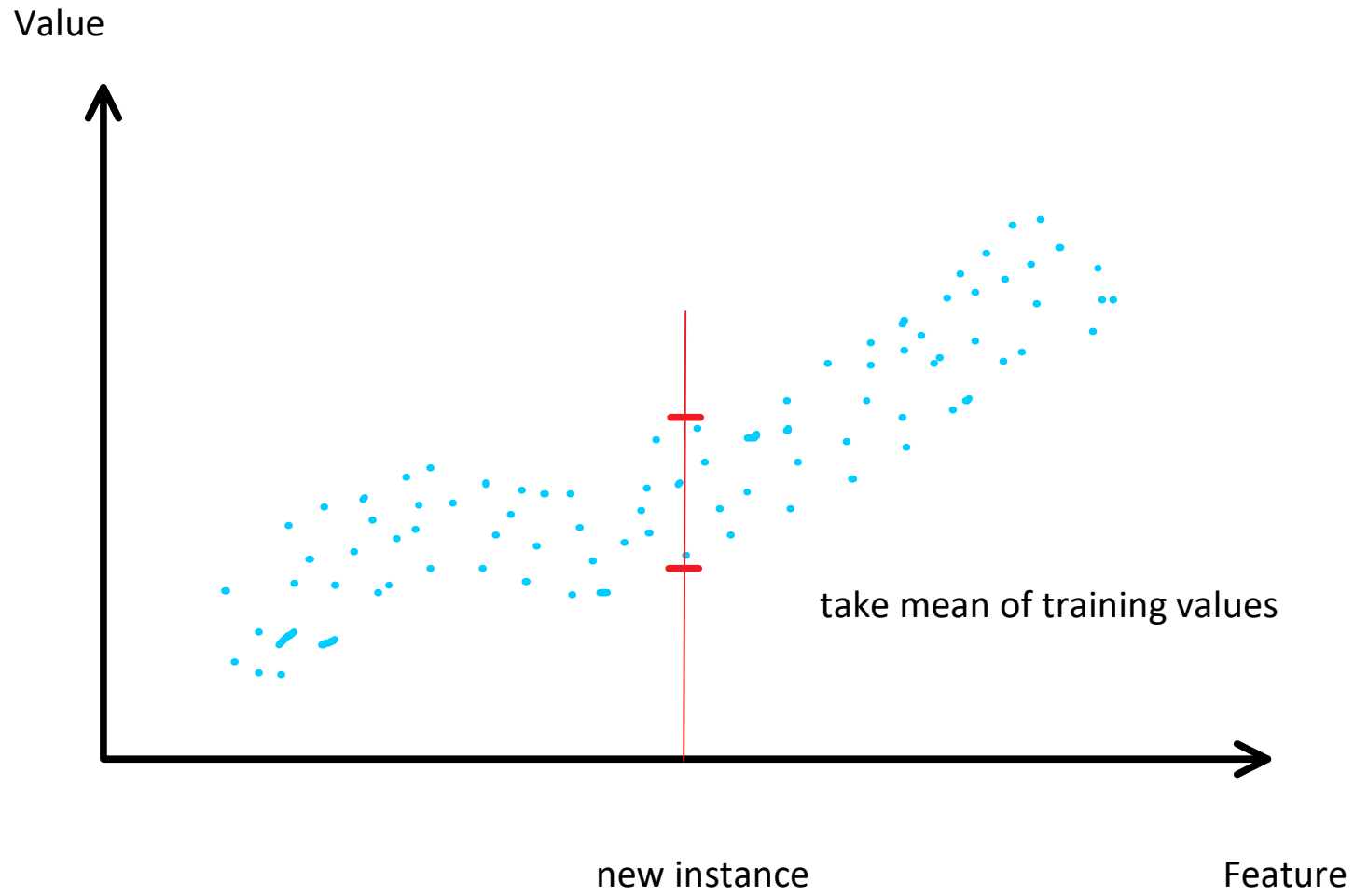
## Data Usage

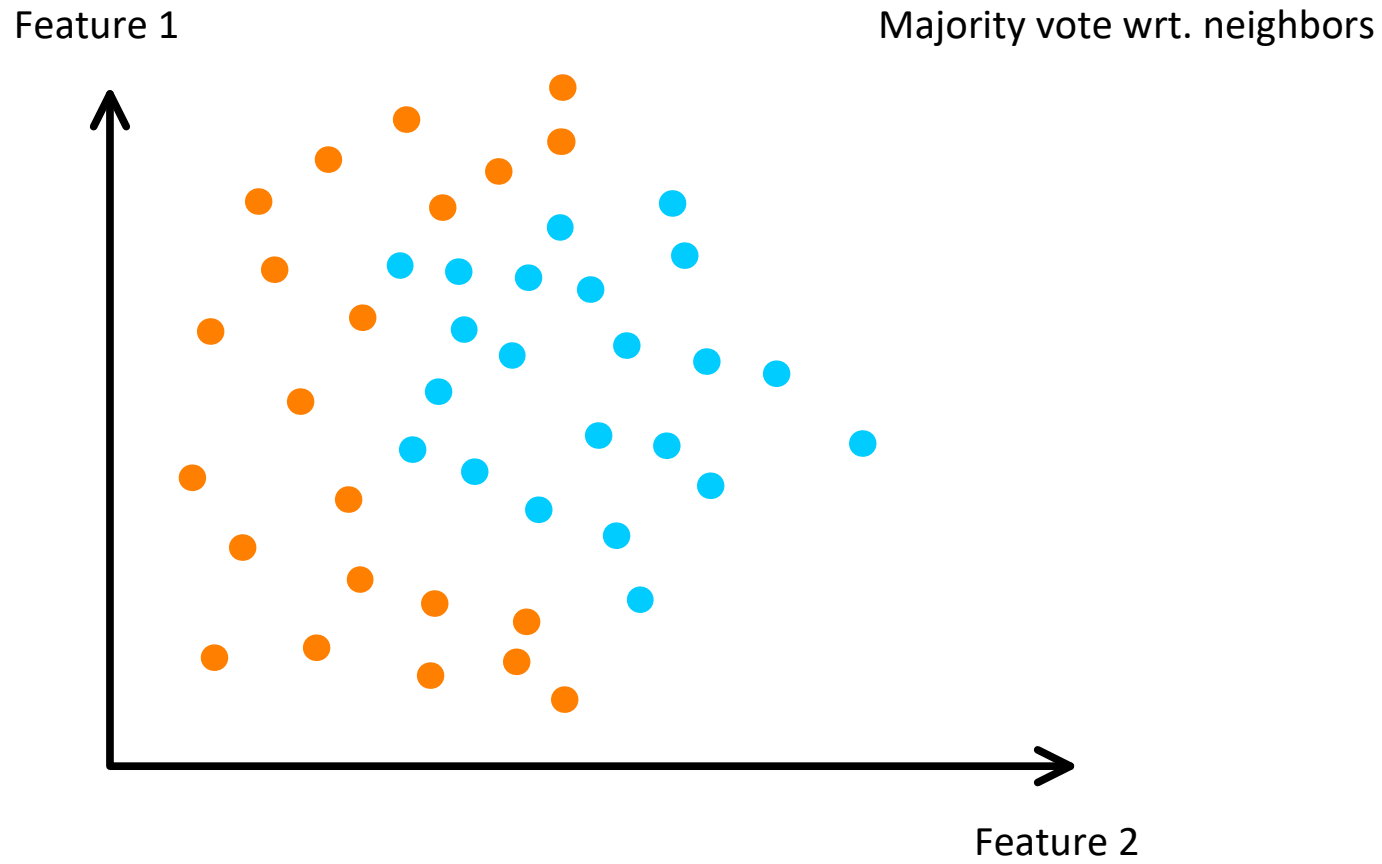
- Batch
- Online

## Method

- Instance-Based
- Model-Based

## Features of instances + target values

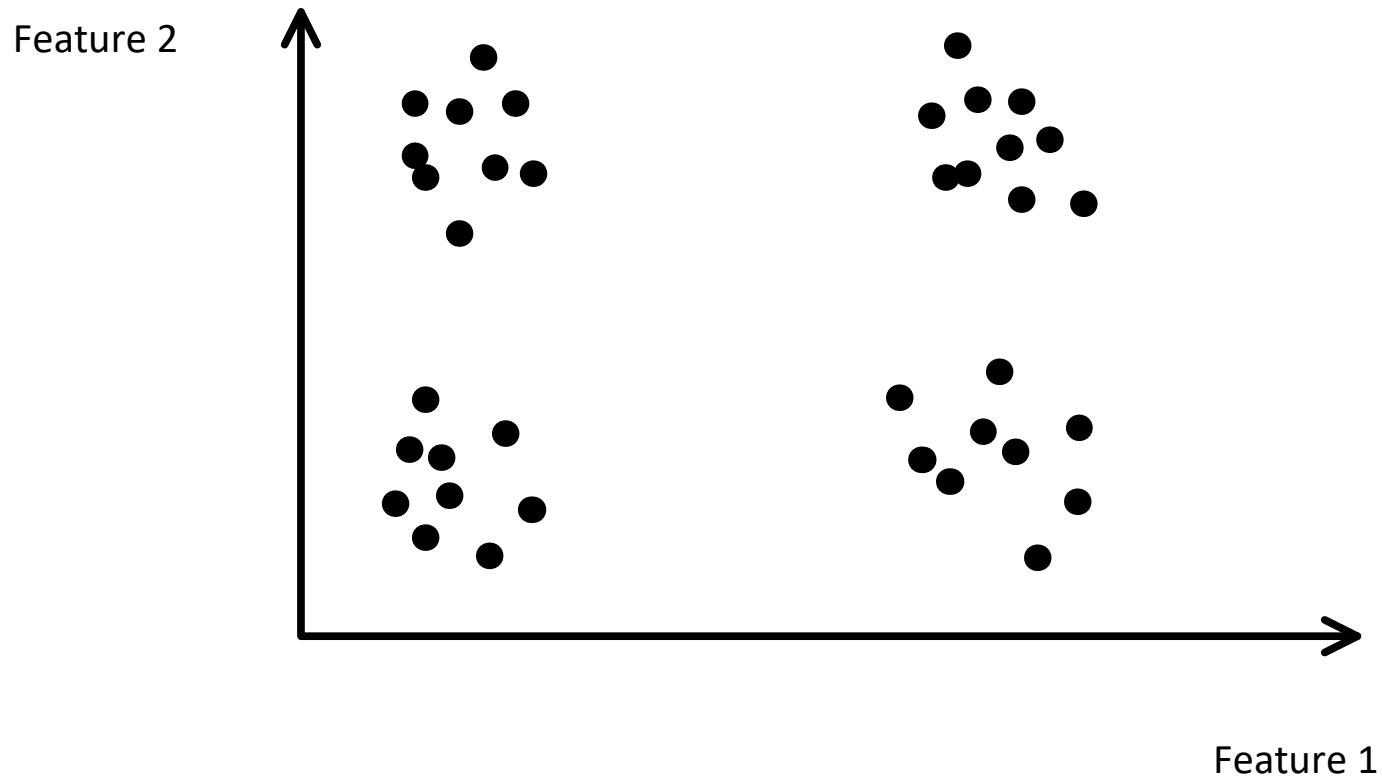




Class tags:

- Class1: orange
- Class2: blue

Features of instances and no class tags

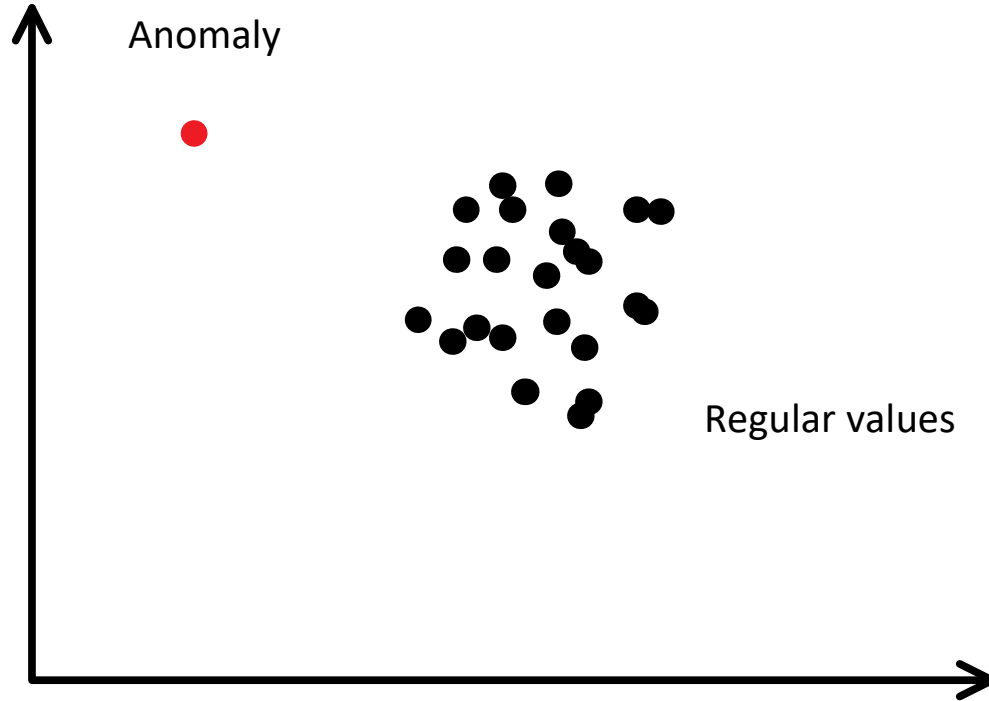


Find sets of data with strong coupling  
and large distance between sets

Features of instances  
Compute cluster of regular instances

Calculate distance

Feature 2



Regular values

Feature 1

## Market Basket Analysis Learning of rules

{ Milk, Bread } -> { Butter }

{ PC, Monitor, Graphics Card } -> { PC Game }

Friday evening

{ Beer } -> { Diapers }



Example: Identification of persons on your private fotos

- Upload of fotos
- System detects same persons on different fotos (unsupervised)
- You tag with name (supervised)

- Agent observes environment
- Chooses action according to strategy
- Executes action
- Gets reward or punishment
- Learns from feedback - adopts strategy
- Repeat until optimal solution found (or at least a satisfying solution)

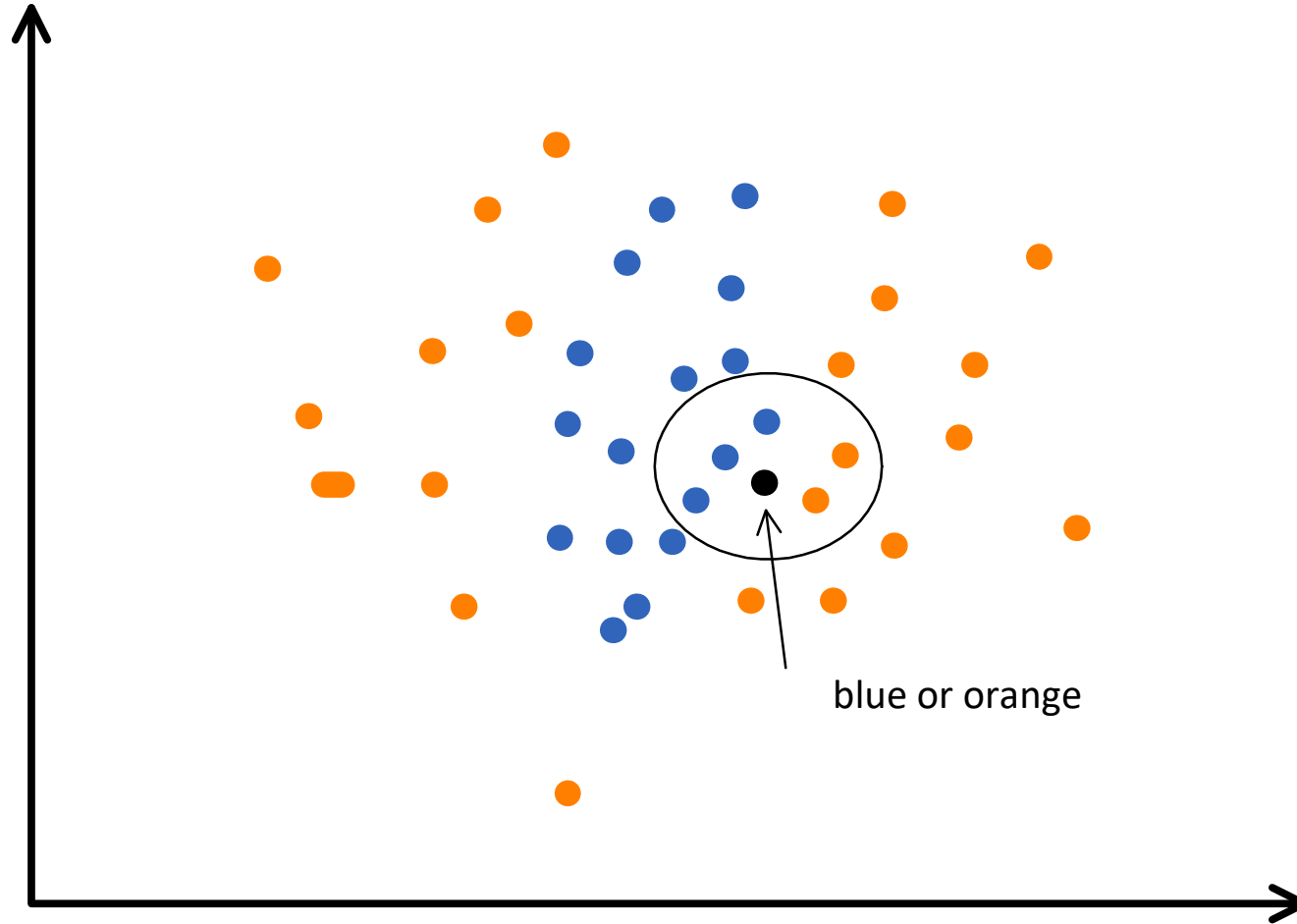
## Batch Learning

- Take all data at once and learn from it
- Complete new take if data changes
- Requires long execution time and resources in case of large training data set

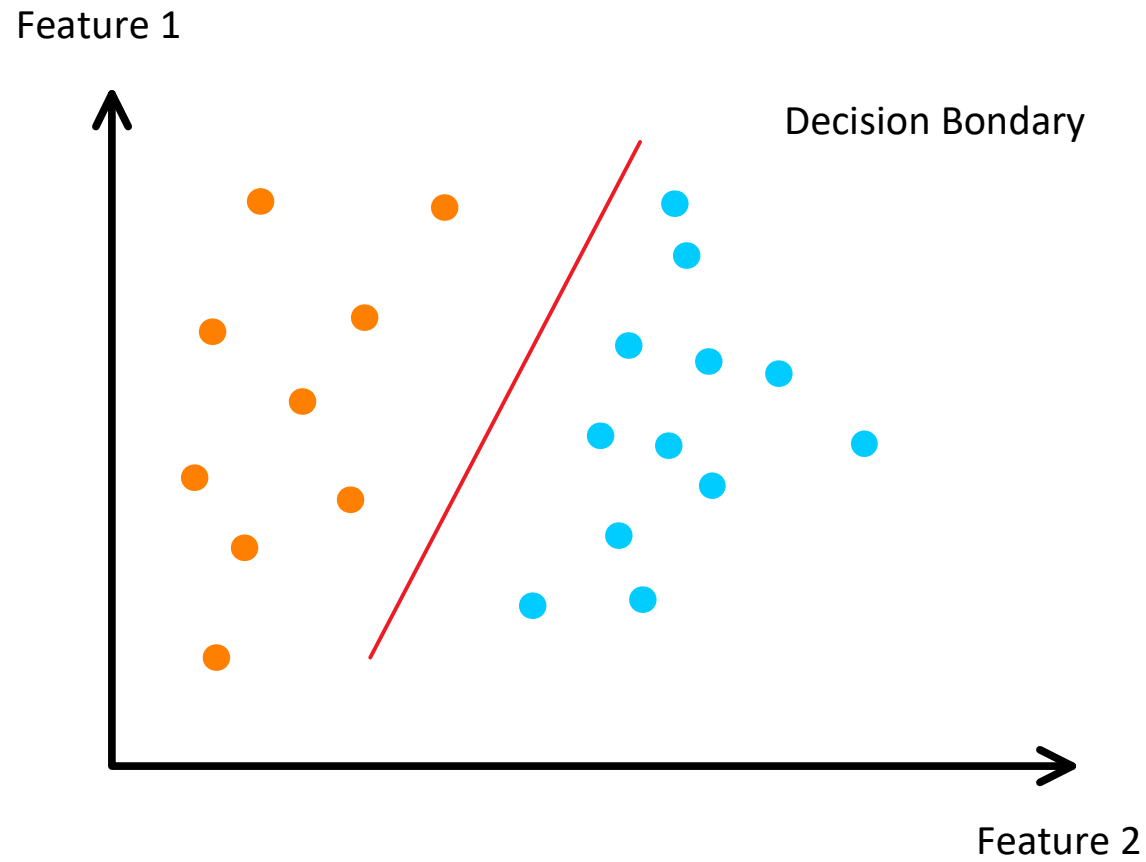
## Online Learning

- Incremental usage of data
- Single data points or minibatches
- Faster adoption to new data
- More volatile models

Similarity to training data



Keep all training data  
Use distance metrics



No storage of trainings data

Just keeping model parameters, e.g. slope and intercept of separating line

Parr, Howard: The Mechanics of Machine Learning, <https://mlbook.explained.ai/>

James, Witten, Hastie, Tibshirani: An Introduction to Statistical Learning, Springer, 2013, <http://www.statlearning.com/>

Géron: Hands-On Machine Learning with Scikit-Learn and TensorFlow, O'Reilly, 2017, <http://shop.oreilly.com/product/0636920052289.do>