

ProtoM teaches you machine learning



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PROTOM



Learn Machine Learning

ProtoM teaches you machine learning. It's simple and interactive so you can get down to business: software that understands speech, Twitter sentiment, stock prices and much, much more.

Learn the basics

Start experimenting

or

Create account

- Start your own projects.
- Save trials automatically and compare your results historically.
- Upload your own data sets for analysis.

Through interactive tutorials

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Tweet Sentiment

Choose a topic!

Let's test the classifier on some live tweets now! Enter a topic in the search box below and we will go grab a bunch of related tweets for you. The tweets will be saved as a new dataset in the *Use test data* menu.

When you have gathered tweets, select the *twitter-<YOUR-SEARCH>* dataset from the *Use test data* menu. Hit *RUN*. The tweets and their predicted labels are shown now. How well did we do? Feel free to keep experimenting with different Twitter topics!

New trial

Trials results

x **Trial-1**

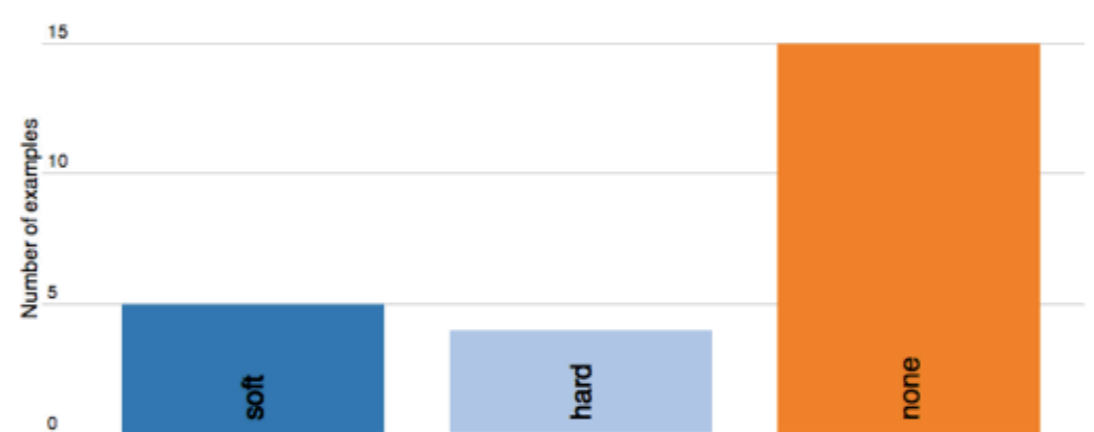
Data **Classifier** **Training and testing**

contact-lenses

Summary

Number of examples: **24** Number of features: **4** Class label: **contact-lenses**

Examples **Features (histograms)**



Class	Number of examples
soft	4
hard	3
none	15

Showing examples from class: soft

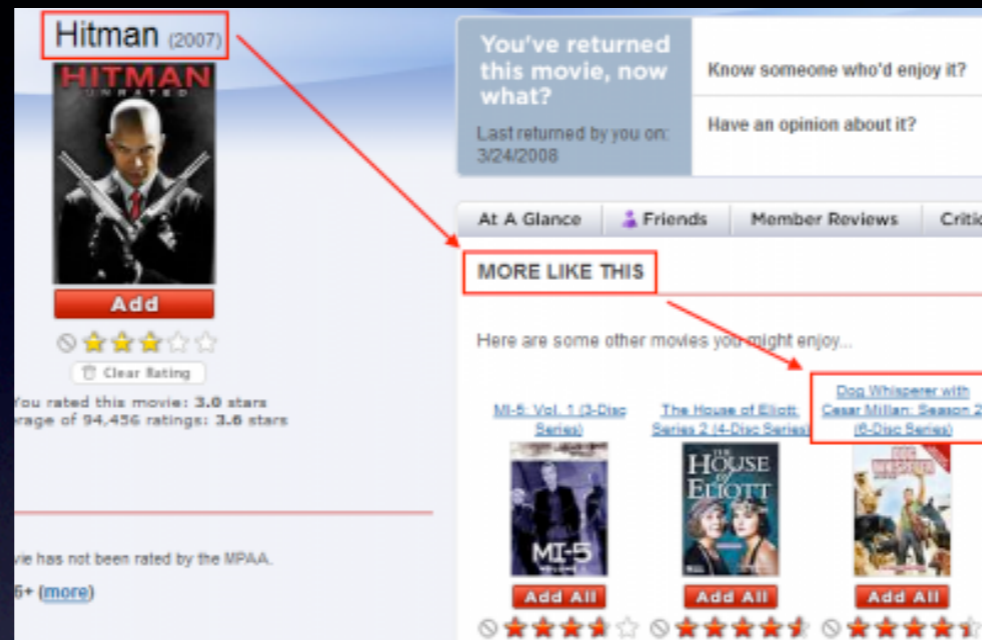
Example 2	
Class	soft
age	young
spectacle-prescrip	myope
astigmatism	no
tear-prod-rate	normal

Example 6	
Class	soft
age	young
spectacle-prescrip	hypermetrope
astigmatism	no
tear-prod-rate	normal

Machine learning's everywhere



Speech recognition



Recommendation



Face recognition

Traditional machine learning course

Finding a decision hyperplane: the perceptron algorithm

Rosenblatt, 1957

Find the linear separator, characterized by parameters θ , that minimizes **training error**:

$$\hat{R}_n(\theta) = \frac{1}{n} \sum_{t=1}^n (1 - \delta(y_t, f(x_t; \theta))) = \frac{1}{n} \sum_{t=1}^n \text{Loss}(y_t, f(x_t; \theta))$$

where $\delta(y, y') = 1$ if $y = y'$ and 0 otherwise.

For simplicity, we use the **zero-one loss** that is 1 for mistakes and 0 otherwise.

Perceptron update rule

- Start with any value for θ (0 is typical)
- Go through training examples x^t, y^t one by one
- Update
 $\theta' \leftarrow \theta + y^t x^t$ if $y^t \neq f(x^t; \theta)$
- Stop when entire training set is categorized correctly

Abstract and theoretical

Whats happening?

Parameter updates tend to correct mistakes

When we make a mistake on x^t ,

- the sign of $\theta \cdot x^t$ disagrees with y^t
- the product $y^t(\theta \cdot x^t)$ is negative
- the updated parameters are $\theta' = \theta + y^t x^t$

Now, if we try to classify x^t with the new parameters, we will have

$$\begin{aligned} y^t(\theta' \cdot x^t) &= y^t(\theta + y^t x^t) \cdot x^t \\ &= y^t(\theta \cdot x^t) + y^{t2}(x^t \cdot x^t) \\ &= y^t(\theta \cdot x^t) + |x^t|^2 \end{aligned}$$

The value of $y^t(\theta \cdot x^t)$ increases as a result of the update (becomes more positive). Successive updates based on x^t will change θ so that it is eventually classified correctly.

What happens when we cycle through the training examples?

Analysis of the perceptron algorithm

- Stops updating parameters when all examples are correctly classified
- Will always find a separator if one exists, in a finite number of updates
- Will not terminate if data are not separable
- Number of updates is related to the **margin**: how close the separating boundary is to the points

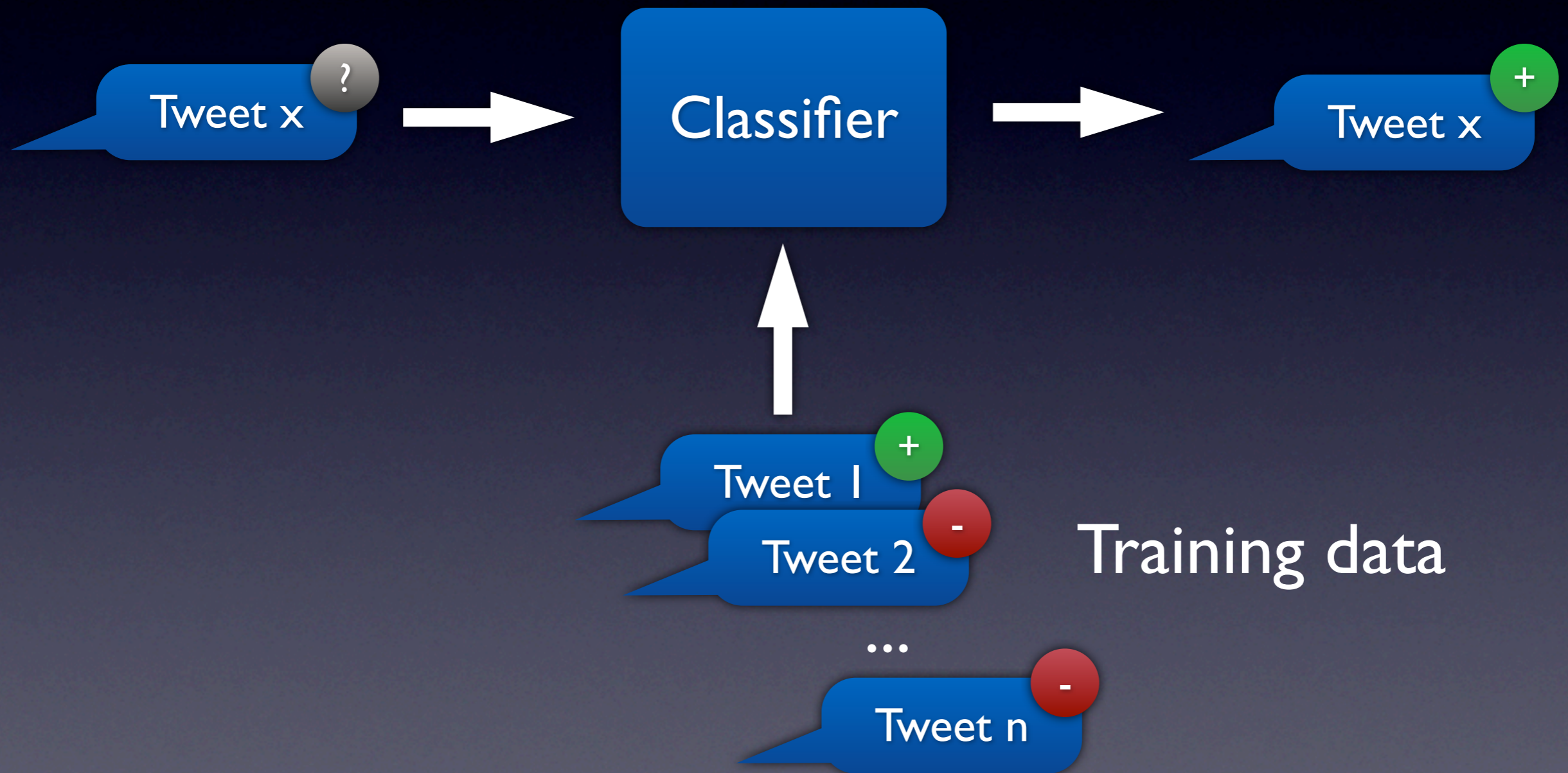
Margin of example i with separator θ^* :

$$\frac{y^i(\theta^* \cdot x^i)}{\|\theta^*\|}$$

Margin of the dataset is the minimum of the margins of the examples

What is ML all about?

An example: Tweet sentiment



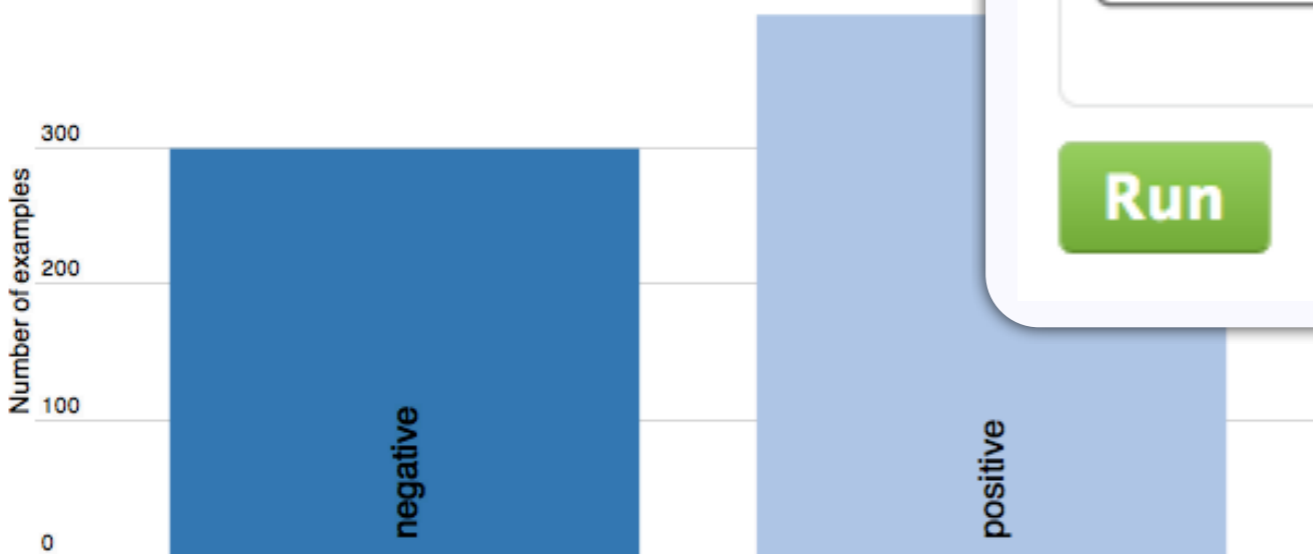
Practical ML with ProtoM

Data **Classifier** **Training and testing**

tweet-sentiment

Summary
Number of examples: 696 Number of features: 1

Examples **Features (histograms)**



Class	Number of examples
negative	~280
positive	~416

Showing examples from class: positive

Example 300	
Class	positive
tweet	why does brissa not even know she has a twitter lauren

Example 301	
Class	positive
tweet	username i thought it was a great love story

Example 302	
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Data **Classifier** **Training and testing**

Cross validation **Use test data**

twitter-love-18-37-15

Run

Practical ML with ProtoM

Predicted results

Predicted classification		
ID	prediction	tweet
1	positive	i love my girlfriend username she is the bestt
2	negative	had the volume up to hear username forgot to turn it down nearly burst my eardrums when i put find your love on that was silly of me
3	positive	username please follow me i love you so so much please i love you
4	positive	i i love you like a love song babyy
5	positive	username love the show any chance of a follow back
6	positive	username hahaha i love how much attitude you have here looks great

ProtoM makes ML accessible and fun



Machine Learning Basics

So, what is machine learning?

It's all about teaching computers to see patterns in data. We see patterns in data all the time. If I show you a picture of a blimp, you'll recognize it as a blimp because you've seen blimps in the past and you've learned that blimps are large, egg-shaped flying machines.

Well, computers have none of this intelligence built-in. Machine learning is all about training software to see data as blimps, cars, people, spoken words, stock prices, and almost anything else you can think of.

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New trial

Trials results

× **Trial-1**

Trial result

Summary

Data:	contact-lenses	Classifier:	rules.ZeroR
Examples:	24	Class:	contact-lenses
Test mode:	Cross validation	Test data:	
Features used:	4 / 4 age spectacle-prescrip astigmatism tear-prod-rate		

Result

Accuracy: **62.5000 %**

Confusion matrix

Classified as =>	1	2	3
soft = 1	0	0	5
hard = 2	0	0	4
none = 3	0	0	15