# Machine Learning

#### for dummies

#### with Python

#### **EUROPYTHON**

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#### Time to leave the office

aooale now next meetina

Search, or say "Ok Google"

nual

## 52 mins to Home

Normal traffic on M6



Navigate / 52 mins via M6

Alternative route / 1 hr 7 min via M6

Tesla autopilot

#### Playing music





#### Your photos organized





Plant

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## Machine Learning is here, it is everywhere and it is going to stay

#### About this presentation

Why Machine Learning (ML) matters

A journey on Machine Learning

Some ML technologies and resources

Some basic ML concepts, with code samples

#### Machine Learning is the next big thing



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#### Are machines already intelligent?



#### Image-net challenge

#### 2015: machines outperform people

ImageNet ñ Release (32326) t life (4486) Personal tion, formation (1) 1112)(176)t (10504) ty, instrumentatior 760) instrument, instru ic device (27) r, adaptor (0) urner (0) nerator (0) aerofoil, control su warning device, ala ce, contraption, cc tor, applier (3)



#### 1997: Deepblue defeats Kasparov



Game of Go

#### 2016: AlphaGo wins world champion Lee Sedol



# TERMINATOR

## The journey

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#### Learning about ML

#### MOOC - Massive Open Online Courses

Contents by the best universities and companies

Udacity, Coursera, EdX

#### Udacity - Intro to Machine Learning

Pattern Recognition for Fun and Profit

- Very well organized contents
- Python + sklearn
- Free
- At your own pace

## Udacity - Intro to Machine Learning Pattern Recognition for Fun and Profit



## Udacity - Intro to Machine Learning <u>Pattern Recognition for Fun and Profit</u>



#### What is Machine Learning?

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# Solving a complex problem features something (data)

prediction

#### First approach, programming

#### tell the computer what to do in very tiny steps

#### First approach, programming

programming does not scale for very complex problems...



#### Machine Learning

#### show the computer some real world data

#### the algorithm will learn from it

#### Machine Learning, implications

### we can train computers to do things we do not know how to do

## CAN YOU ELABORATETHATP

#### ML example: character recognition



Not-MNIST dataset Thousands of 28x28 grayscale images with labels

#### ML step 1: get samples (training data) $\mathbf{0}$ $\mathbf{O} \mathbf{O} \mathbf{O} \mathbf{O}$ $\mathbf{O} \mathbf{O} \mathbf{O} \mathbf{O}$ $\mathbf{O} \mathbf{O} \mathbf{O} \mathbf{O}$ FGFJ $\mathbf{O} \mathbf{O} \mathbf{O} \mathbf{O}$ $\mathbf{0}$ $\bigcirc \bigcirc \bigcirc \bigcirc \bigcirc$ $\bigcirc \bigcirc \bigcirc \bigcirc \bigcirc$ features x 1000s

labels x 1000s

#### ML step 2: choose an algorithm



Linear regression Support Vector Mach. k-Means **Decision** Trees Random Forests Neural networks Convolutional NN Naive Bayes

#### ML step 3: train your algorithm

features x 1000s ML algorithm

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labels x 1000s

FGFJ

#### ML, last step: getting predictions

features (data)



ML algorithm

prediction

#### Tricky Question

# How good are our predictions?



## The Tools

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#### The Tools: Python

• Opensource

Expressive

Interpreted, dynamically typed

• Widely used many different problems

• Batteries included: Notebook, Libraries

#### The Tools: sklearn

earn.

- Opensource, Python
- Wonderful documentation
  - Support to full ML lifecycle:
    - Feature engineering
    - Algorithms
    - Validation
    - Datasets

#### A summary of ML process

- Get features (with labels)
- Choose and configure an algorithm
- Train your algorithm
- Do predictions
- Validate your results

#### train your model

train_ds	test_ds							

clf = LogisticRegression()
clf.fit(tr\_ds, tr\_lbl) # fit with train dataset and train labels

#### make predictions

pred = clf.predict(test\_dataset)

#### How good are our predictions?



test\_predicions = clf.predict(test\_dataset)
acc = accuracy\_score(test\_labels, test\_predictions)

## 89% accuracy



#### Improving prediction results

#### Training data

#### Algorithm + config



## THE TWO STATES OF EVERY PROGRAMMER



## I AM A GOD.



### I HAVE NO IDEA WHAT I'M DOING.

#### Udacity - Deep Learning

#### Take machine Learning to the next level

ML branch based on algorithms that use multiple processing layers

- By Google
- Python and Tensorflow
- No wine for the moment :-(

#### The Tools: TensorFlow

#### • Opensource, Python

• Deep Learning

Data flow graphs.
 Nodes: mathematical operations
 Edges: Tensors, multidimensional arrays

#### Simplest Neural Network

LAYER 1

LAYER 2



#### Deep Learning as a chain of operations

11\_logits = tf.matmul(tf\_test\_dataset, l1\_weights) + l1\_biases
11\_output = tf.nn.relu(l1\_logits)
12\_logits = tf.matmul(l1\_output, l2\_weights) + l2\_biases
test prediction = tf.nn.softmax(l2 logits)



# Let's recap

















#### Thank you for your attention