

Codettes & Geekettes Code Pub Munich

Data Science Workshop



```
if _operation == "MIRROR_X":  
    mirror_mod.use_x = False  
    mirror_mod.use_y = True  
    mirror_mod.use_z = False  
elif _operation == "MIRROR_Z":  
    mirror_mod.use_x = False  
    mirror_mod.use_y = False  
    mirror_mod.use_z = True  
  
#selection at the end -add back the deselected mirror modifier object  
mirror_ob.select= 1  
modifier_ob.select=1  
bpy.context.scene.objects.active = modifier_ob  
print("Selected" + str(modifier_ob)) # modifier ob is the active ob  
#mirror_ob.select = 0  
base = bpy.context.selected_objects[0]  
base.select = True
```

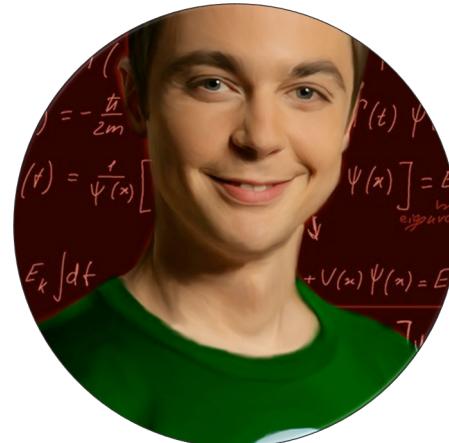


Dr. Lydia Nemec



Data Scientist @ Zeiss
Theoretical Physicist by training

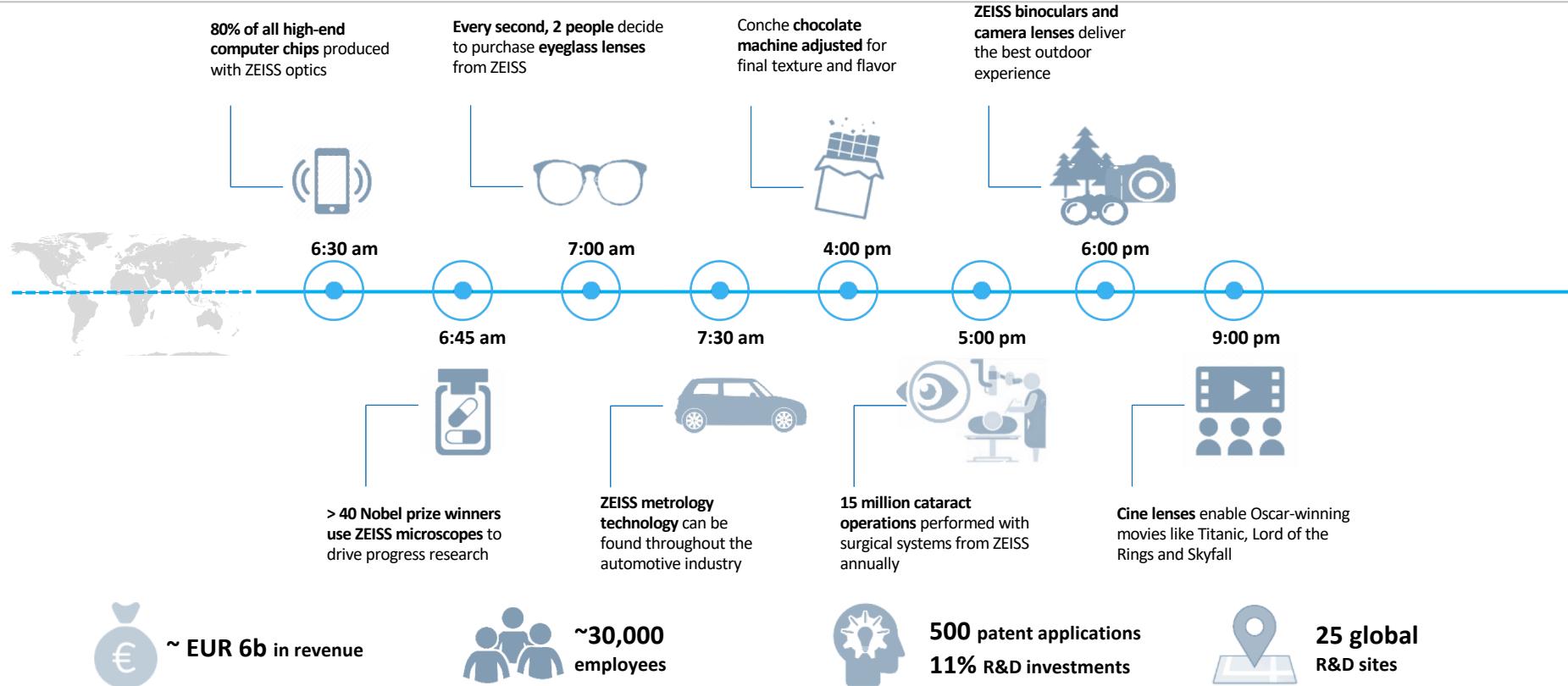
but I come in a
wrapping similar to
Penny



I am a bit like
Dr. Sheldon Cooper,

One day without ZEISS?

A “user journey”





ARTWORK: TAMAR COHEN, ANDREW J BUBOLTZ, 2011, SILK SCREEN ON A PAGE FROM A HIGH SCHOOL YEARBOOK, 8.5" X 12"

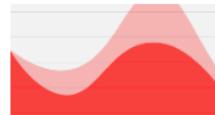
DATA

Data Scientist: The Sexiest Job of the 21st Century

by Thomas H. Davenport and D.J. Patil

FROM THE OCTOBER 2012 ISSUE

WHAT TO READ NEXT



Visualizations That Really Work

Data Scientists apply numerical methods like Machine Learning to extract insights from data.

Math, Numeric & Statistic

- Machine Learning (AI)
- Statistical modelling
- Linear Algebra & Optimization

The Scientific Mind

- Logical & independent mind
- Planning, conducting & evaluate experiments
- Excellent analytical skills
- Meticulous attention to quality and accuracy



Computer Science & Programming

- Software development
- Programming Language (e.g. python)
- Databases (SQL/ No-SQL)
- Cloud Computing

Communication, Soft Skills & Visualisation

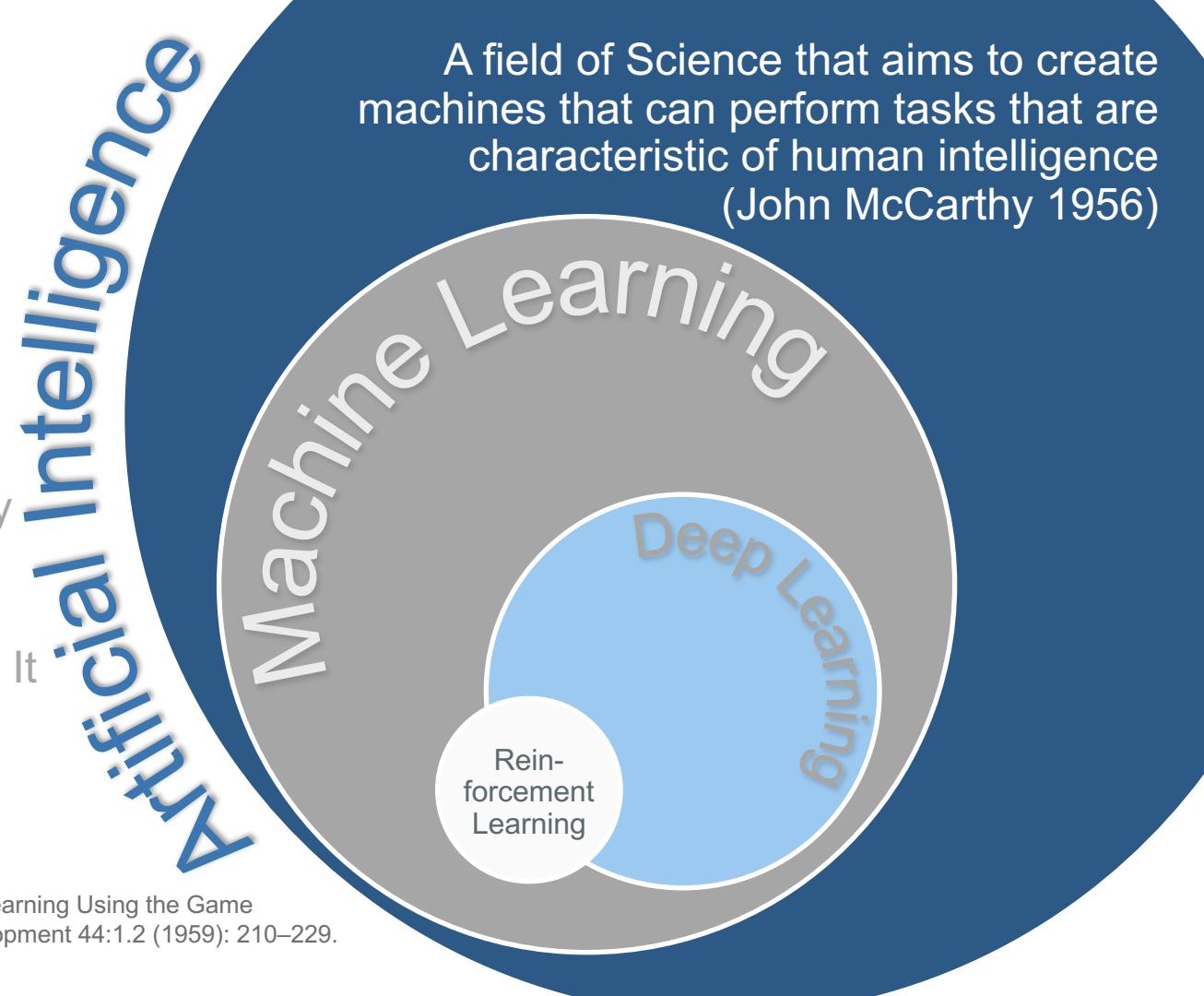
- Collaborative, strategic, proactive, creative and innovative
- Influence without authority
- Translate data-driven insights into impactful decisions and actions
- Data Visualisation

Data Science
combines the complexity of
Software Development,
the challenges of applied **numerical analysis**
with the additional **dynamic** introduced by **data!**

Machine Learning refers to a set of algorithms that allow computers to learn from data without being explicitly programmed. [1]

Deep Learning is part of a broader family of machine learning methods based on artificial neural networks. It belongs to the class of hierarchical learning algorithm.

A field of Science that aims to create machines that can perform tasks that are characteristic of human intelligence
(John McCarthy 1956)



[1] Samuel, Arthur L. „Some Studies in Machine Learning Using the Game of Checkers,“ IBM Journal of Research and Development 44:1.2 (1959): 210–229.

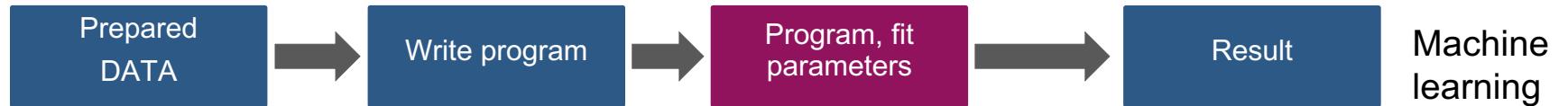
Machine Learning: A different way of software development



Traditional Software

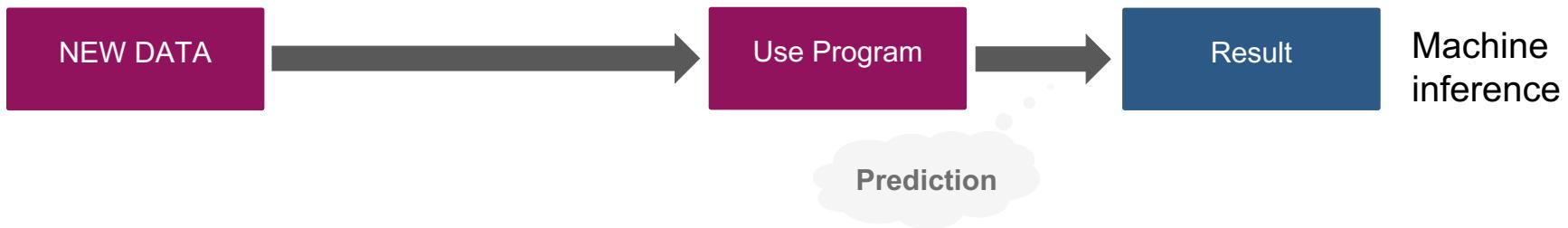


Machine Learning Software

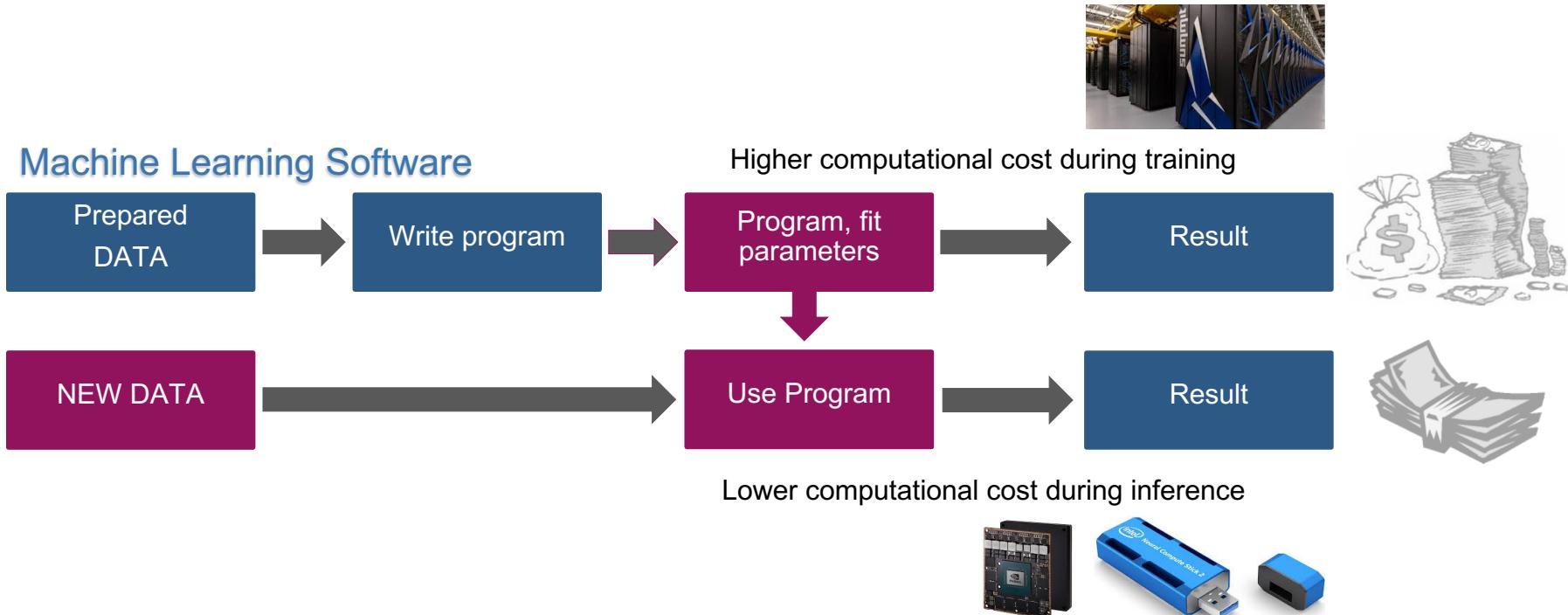


Prediction

Prediction
is the process of filling in missing information!
Prediction takes information you have, often called "data",
and uses it to extract information you need.

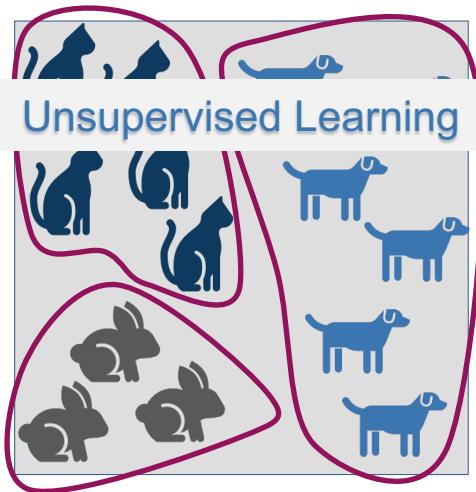


Machine Learning cost asymmetry in training and inference

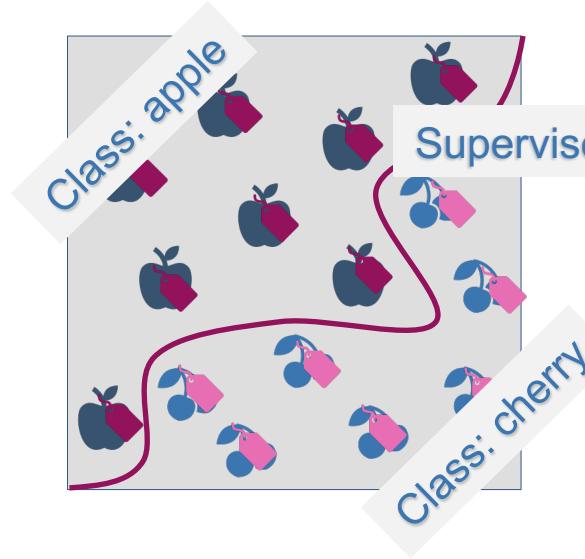


Summit Oak Ridge National Lab since November 2018 fastest supercomputer in the world

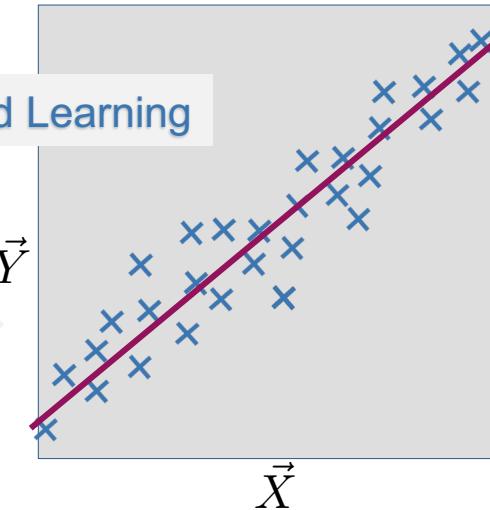
Clustering



Classification



Regression



- Clustering algorithm try to find structure or pattern in uncategorized data.
- Clustering algorithm map input data to output labels.
- Regression algorithm map input data to a continuous output value.

Classification and regression model development needs labelled input data.

Unsupervised Learning

- In unsupervised learning model, only input data is given.
- Uses unlabelled data
- Number of classes is not known.
- Less accurate and trustworthy method.
- Difficult to validate the reliability of the Machine Learning model.
- Algorithm examples:** Cluster algorithms (K-means, Hierarchical clustering)
Signal separation (Principal & Independent component analysis)
Neural Networks (autoencoders),
anomaly detection (isolation forest).

Supervised Learning

- In a supervised learning model, input and output variables are given.
- Uses labelled data
- Learning a link between the input and the outputs.
- Possible to develop highly accurate and trustworthy method.
- Difficult to get a sufficient amount of high quality labelled data.
- Algorithm examples:** Support vector machine, (deep) neural network, Linear and logistics regression, random forest, and Classification trees (forests)

Linear Regression; the simplest example of a Machine Learning Algorithm

$$\vec{Y} \approx f(\vec{X}; \{p_1, p_2, \dots, p_N\})$$

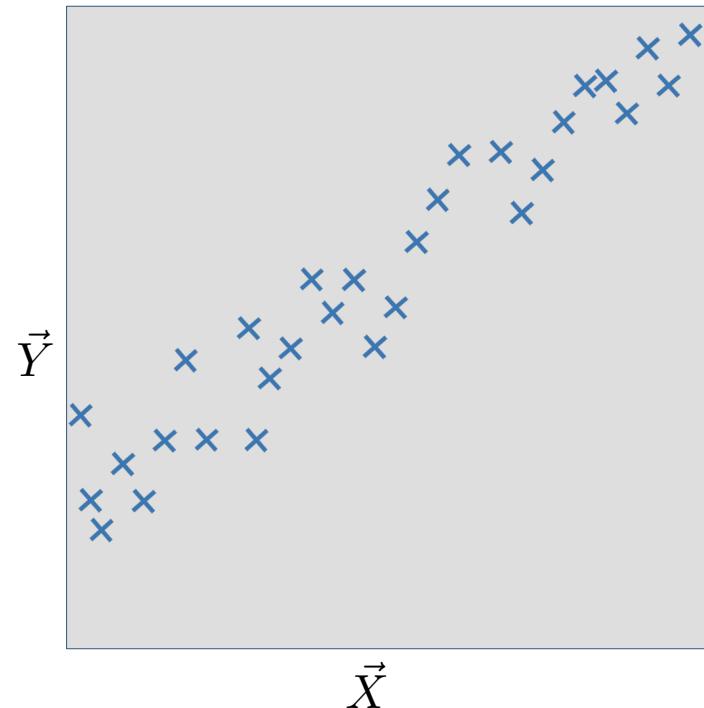
- The function f describes the relationship between \vec{X} and \vec{Y} .
- In general, f is not known.
- \vec{X} and \vec{Y} can be multidimensional vectors

Example:

x duration of a taxi drive in seconds
y price of the Taxi drive

Example multidimensional Input:

\vec{X} (x_1) duration and (x_2) distance of a taxi drive
y price of the Taxi drive



Linear Regression; the simplest example of a Machine Learning Algorithm

$$\vec{Y} \approx f(\vec{X}; \{p_1, p_2, \dots, p_N\})$$

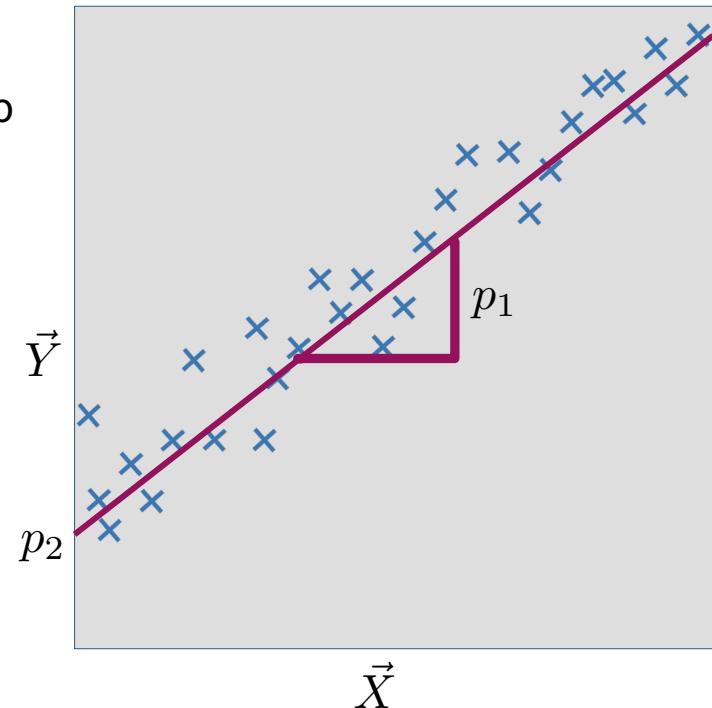
- The simplest form of f approximating the relationship between \vec{X} and \vec{Y} is a straight line.
- $\{p_1, p_2, \dots, p_N\}$ are the parameters determining the function f

$$Y = f(X; p_1, p_2)$$

$$N = 2$$

$$Y = p_1 X + p_2$$

- p_1 slope and p_2 Y-intersection of the linear equation
- The equation is fully defined by p_1, p_2



Linear Regression; the simplest example of a Machine Learning Algorithm

$$\vec{Y} \approx f(\vec{X}; \{p_1, p_2, \dots, p_N\})$$

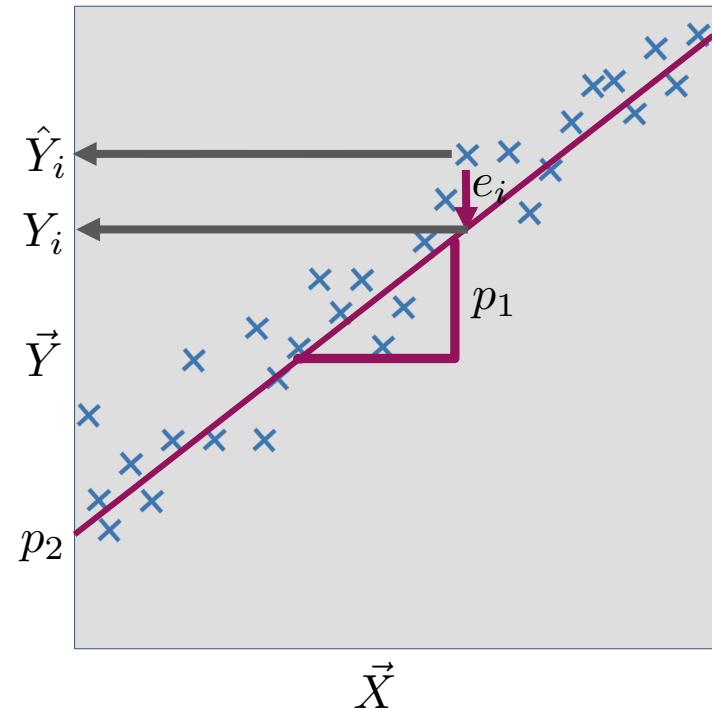
- Error e_i between the function value \hat{Y}_i and the data point \hat{Y}_i

$$e_i = |\hat{Y}_i - Y_i|^2$$

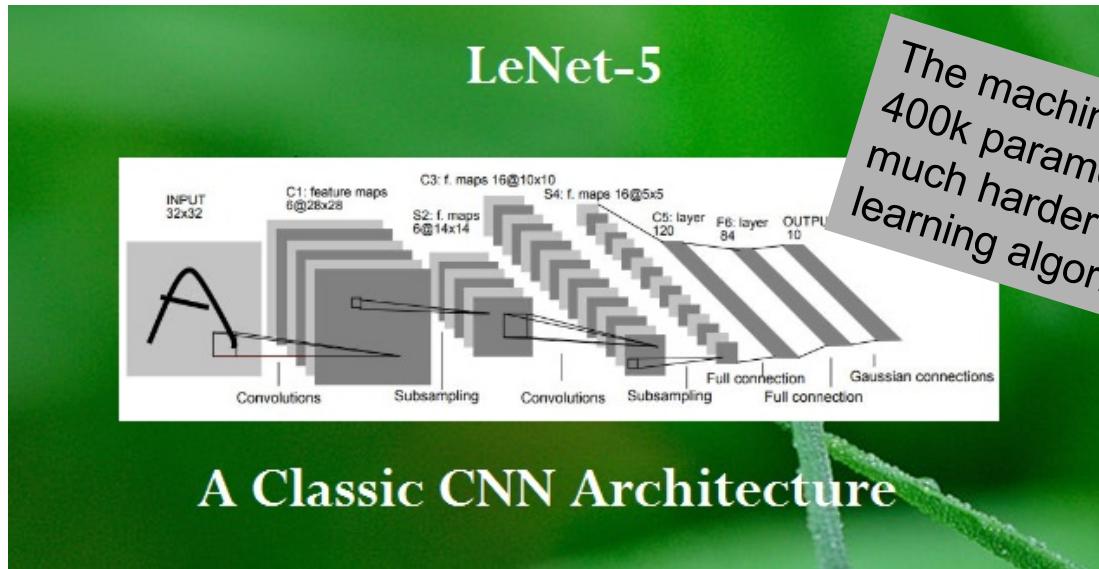
- The machine Learning part is finding p_1, p_2
- p_1, p_2 is determined by minimising the sum of e_i

$$MSE = \frac{1}{N} \sum_{i=1}^N |\hat{Y}_i - Y_i|^2$$

- Using gradient descent: updating p_1, p_2 to reduce the cost function (MSE).



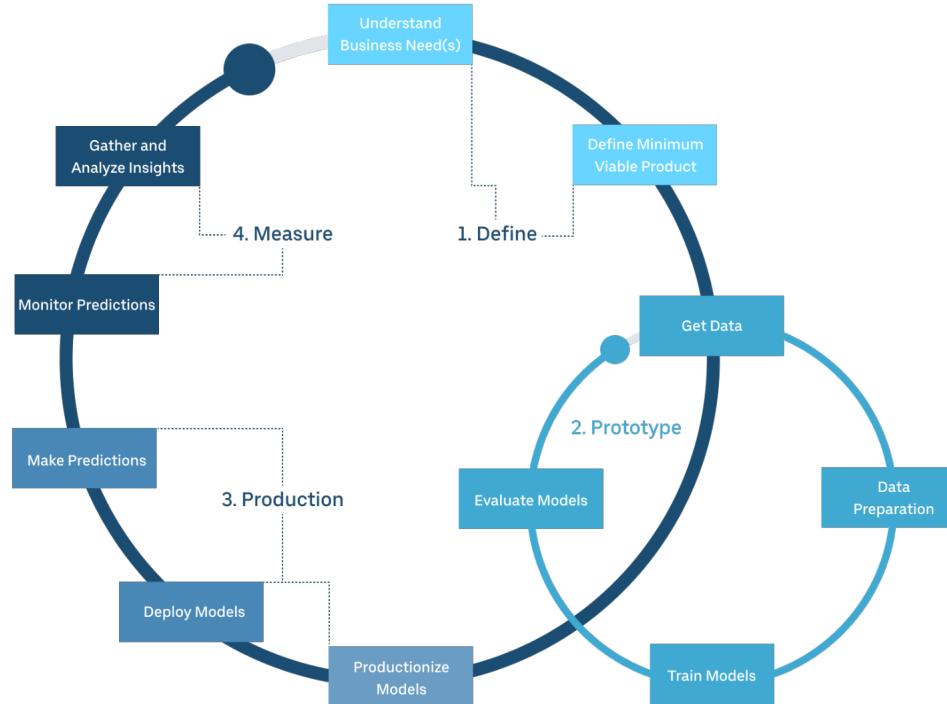
Example of a Machine Learning Algorithm: Deep neural network for computer vision



The machine Learning part of fitting the 400k parameter $\{p_1, p_2, \dots, p_{388272}\}$ is much harder in advanced machine learning algorithm like deep neural nets.

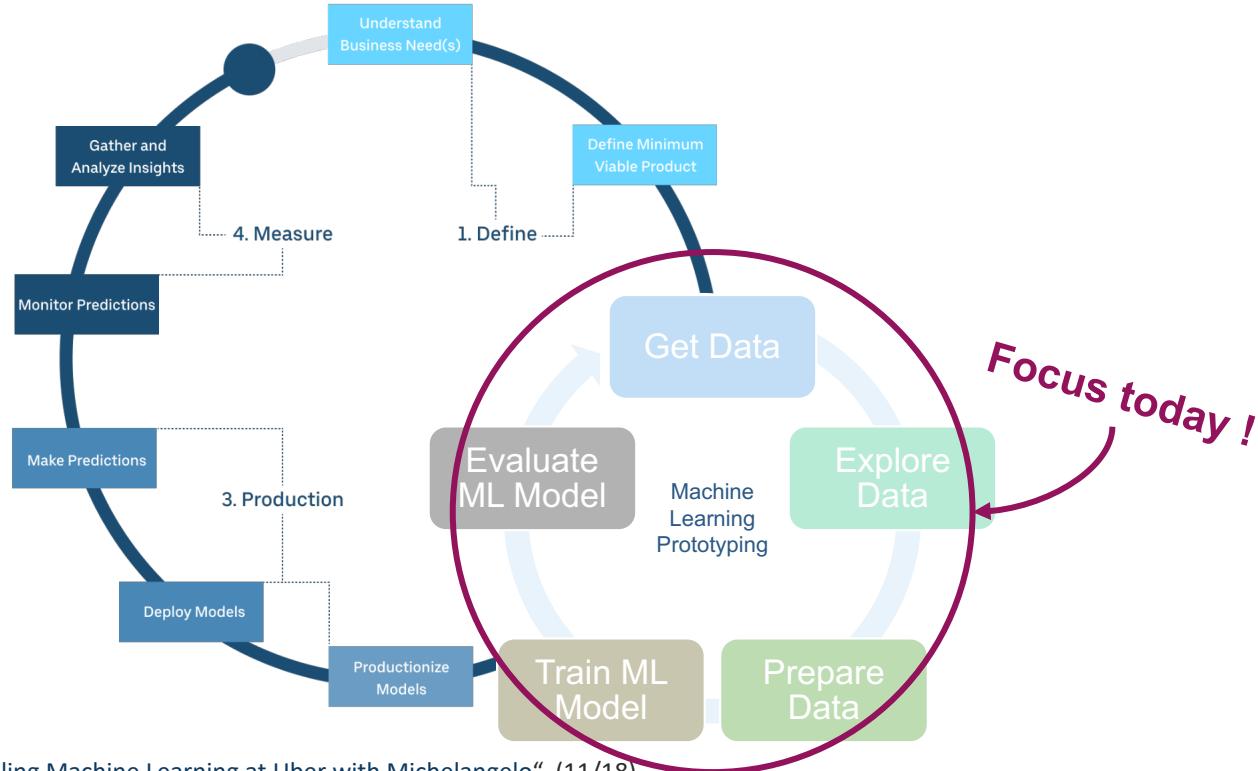
- LeNet-5 is a typical example of a deep neural network for computer vision for a 32×32 pixel image the Net contains **~400k parameters** to optimise.

Data Science Projects: A Highly Iterative Process



[1] Jeremy Hermann and Mike Del Balso; „[Scaling Machine Learning at Uber with Michelangelo](#)“ (11/18)

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Data Science: The surrounding Infrastructure is Vast and Complex

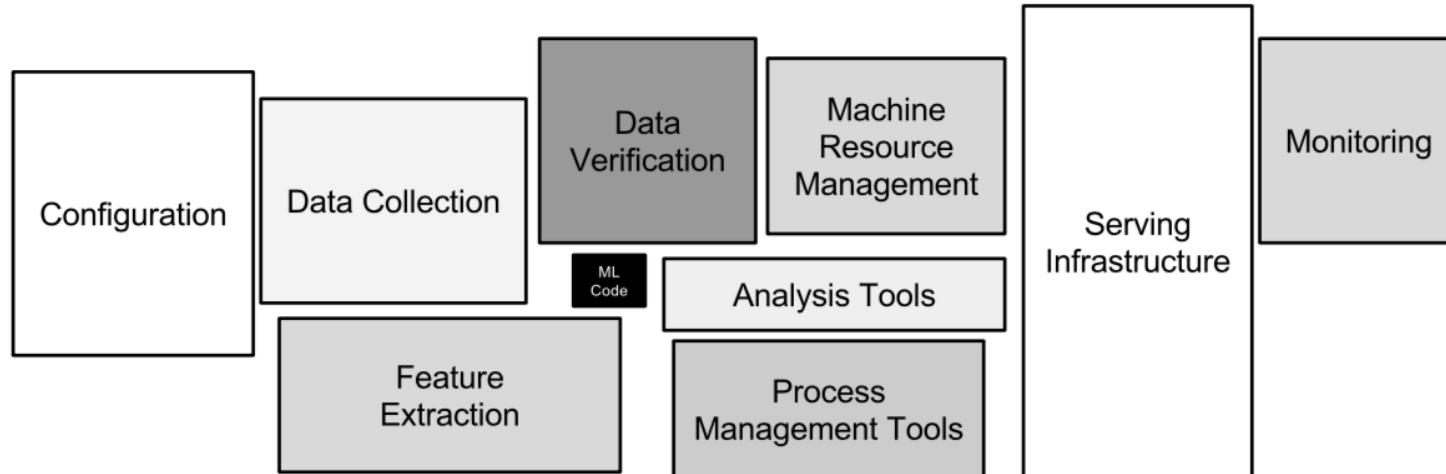


Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small black box in the middle. The required surrounding infrastructure is vast and complex.

The Use Case today New York City Taxi



The Use Case today New York City Taxi





Dr. Lydia Nemec
Data Scientist @ Zeiss



<https://www.linkedin.com/in/lydianemec/>
[@LydiaNemec](https://twitter.com/LydiaNemec)

Thank you for your attention

