

Dr. Michael Vogt · 6. Dezember 2018 · Uni Siegen

# Deep Learning – Die Revolution der künstlichen Intelligenz

**smiths detection**  
bringing technology to life

# Künstliche Intelligenz in den Medien

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## **Google AI algorithm masters ancient game of Go**

Deep-learning software defeats human professional for first time.

“Deep learning is killing every problem in AI.”

The AI Revolution: Why Deep Learning Is Suddenly Changing Your Life

## **The First Person to Hack the iPhone Built a Self-Driving Car**

George Hotz is taking on Google and Tesla by himself.

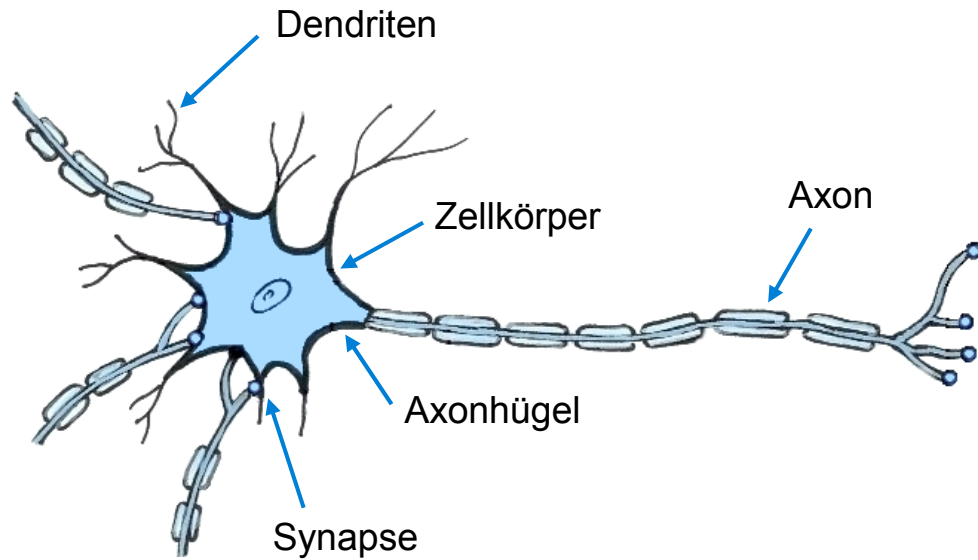
“Deep learning was like  
ten breakthroughs at once –  
and the best is still to come.”

## **You're killing people!**

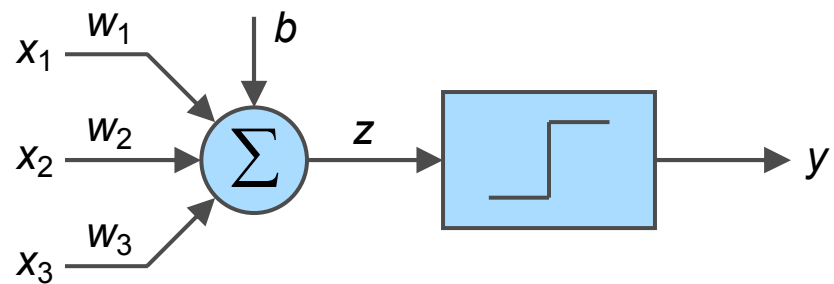
Elon Musk attacks critics of self-driving cars

“The cost of a world-class deep learning expert was  
about the same as a top NFL quarterback prospect”

# Neuronen



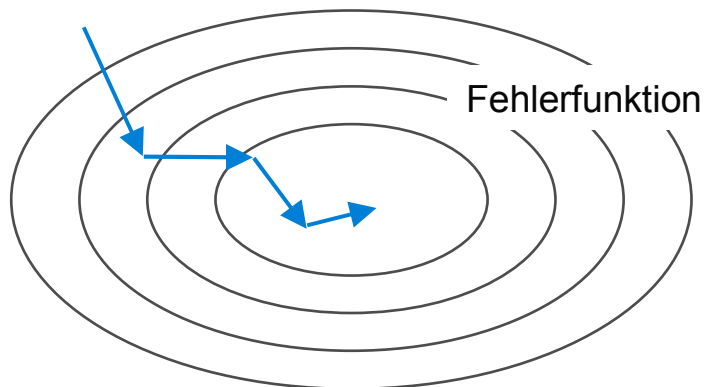
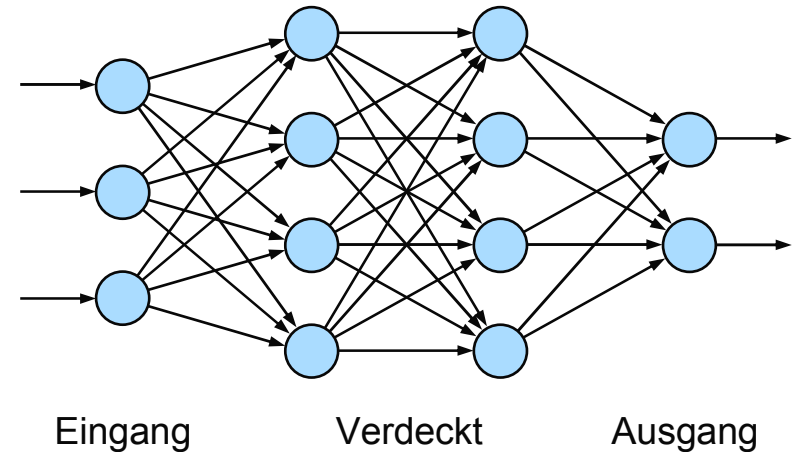
## McCulloch-Pitts-Modell (1943)



# Künstliche neuronale Netze

## Feed-Forward-Netze

- Nur Vorwärts-Verbindungen
- Perzeptron (Rosenblatt, 1958)
- „Eine Schicht ist genug!“ (1989)



## Trainieren von Netzen

- Lernen aus empirischen Daten
- Error Backpropagation & Co.  
(→ Gradientenabstieg)

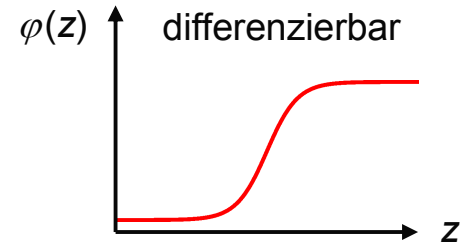
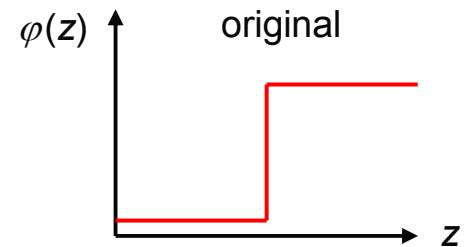
# Überwachtes Lernen

Aufgabe: Approximiere  $f_{\mathbf{w}}(\mathbf{x})$  auf  $\{(\mathbf{x}_k, y_k)\}_{k=1}^N$

- Minimiere Fehler  $E = \sum_k (y_k - f_{\mathbf{w}}(\mathbf{x}_k))^2$
- Gradient aus Error-Backpropagation

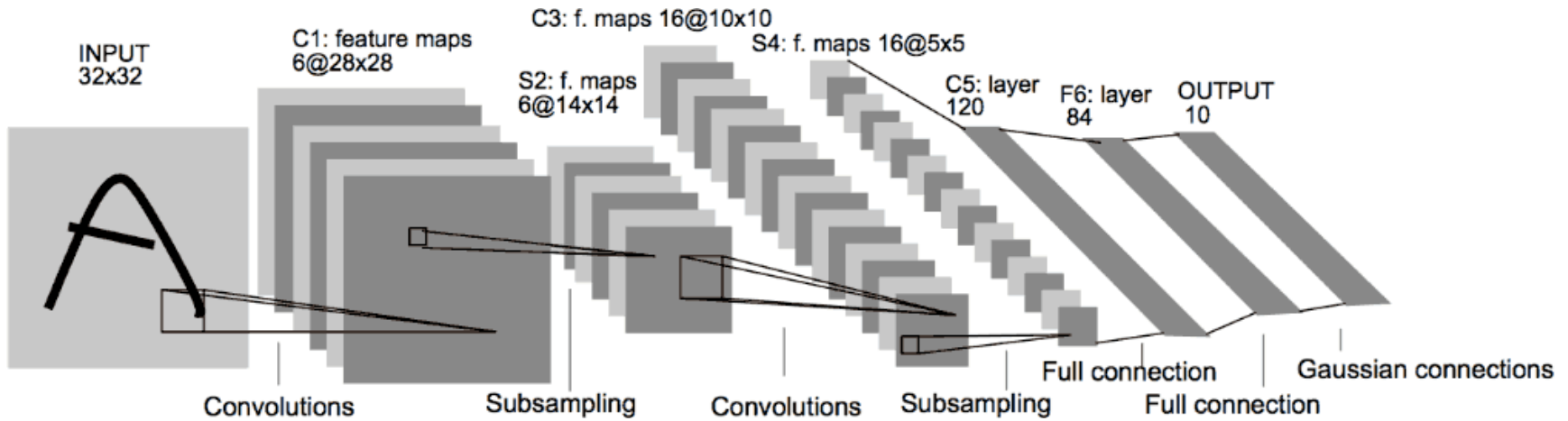
$$\frac{\partial E}{\partial w_{ij}} = \frac{\partial E}{\partial f_{\mathbf{w}}} \cdot \underbrace{\frac{\partial f_{\mathbf{w}}}{\partial z} \cdot \frac{\partial z}{\dots}}_{= \varphi'(z)} \cdot \frac{\dots}{\partial w_{ij}}$$

- Parameter-Update:  $w_{ij}^{\text{neu}} = w_{ij}^{\text{alt}} - \eta \cdot \frac{\partial E^{\text{alt}}}{\partial w_{ij}}$



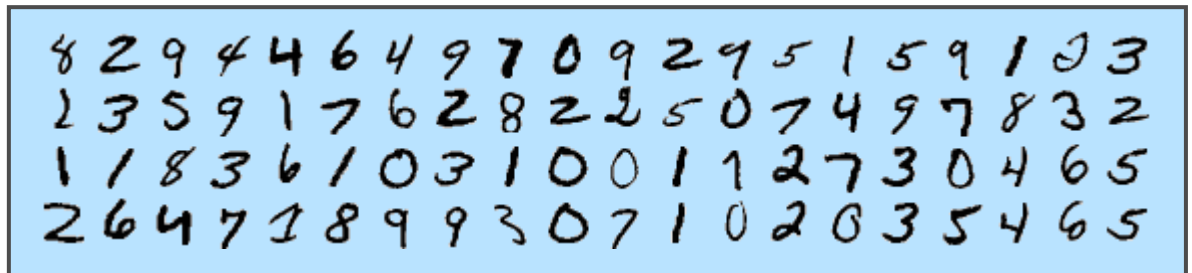
Maßnahmen: Momentum, Regularisierung, Early-Stopping, ...

# LeNet (1989)

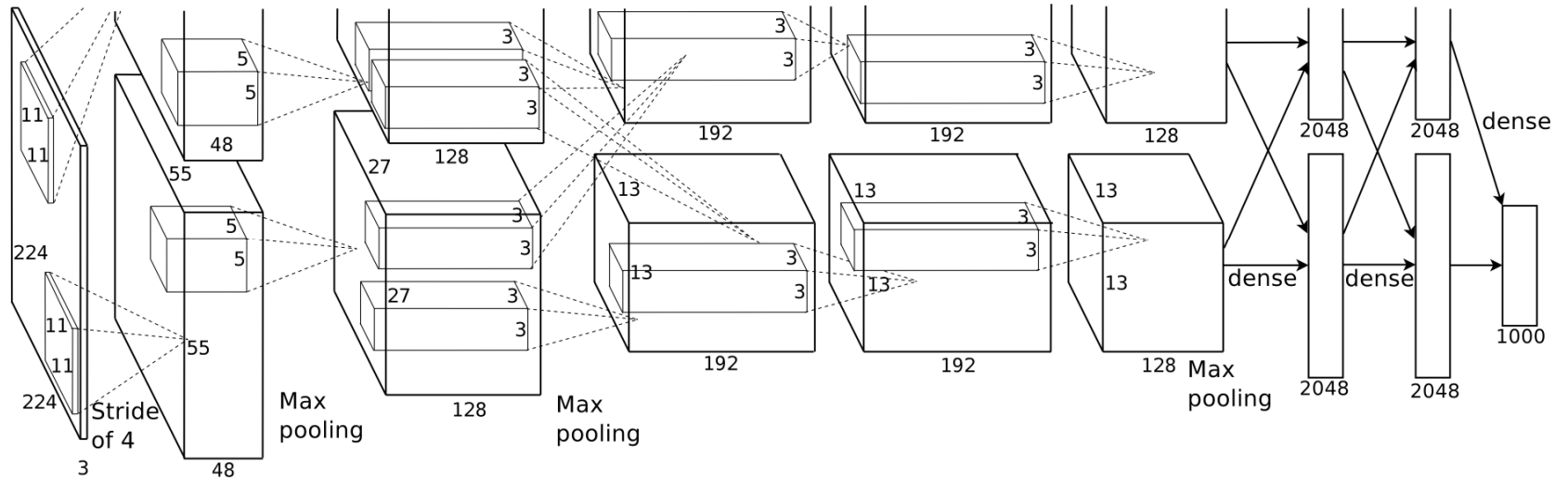


LeCun et al. (1998)

MNIST-Datensatz:  
Handgeschriebene  
Ziffern



# AlexNet (2012)



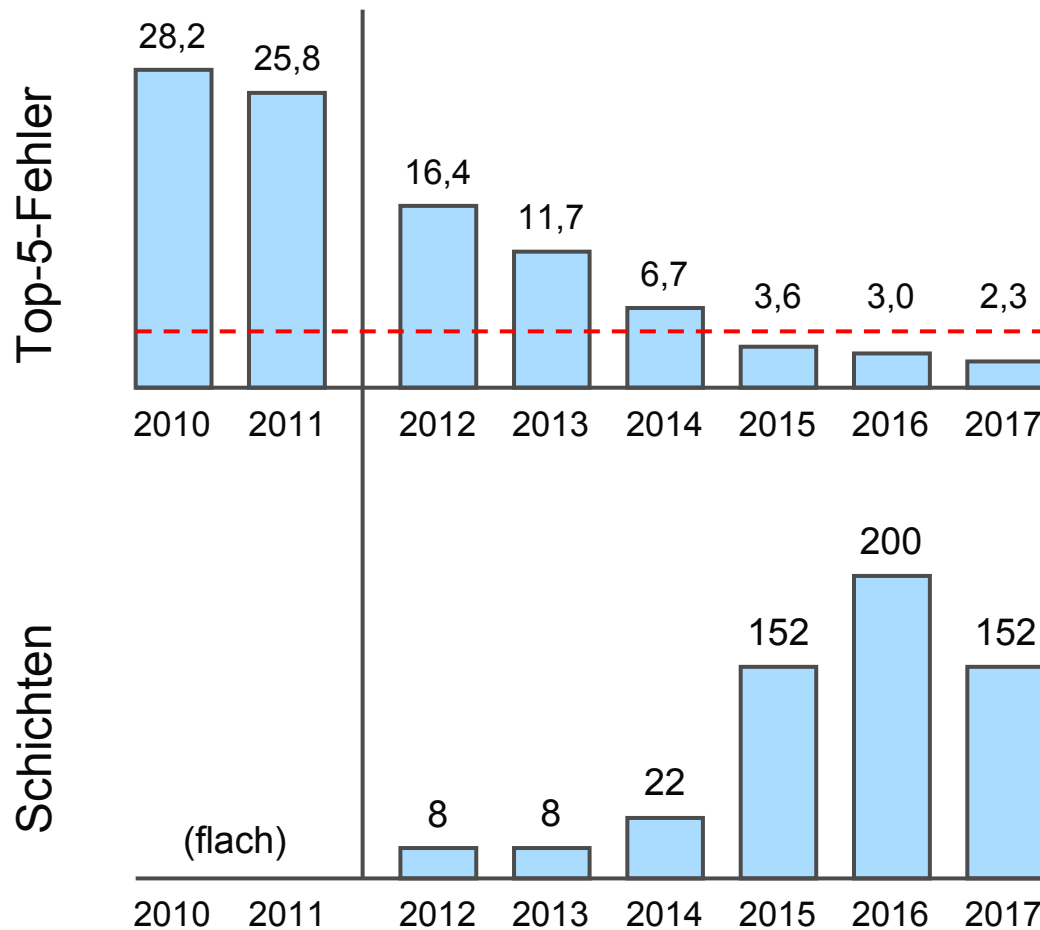
IMAGENET



grille	mushroom	cherry	Madagascar cat
convertible	agaric	dalmatian	squirrel monkey
grille	mushroom	grape	spider monkey
pickup	jelly fungus	elderberry	titi
beach wagon	gill fungus	ffordshire bulterrier	indri
fire engine	dead-man's-fingers	currant	howler monkey

<http://www.image-net.org/>

# ImageNet Challenge (ILSVRC)

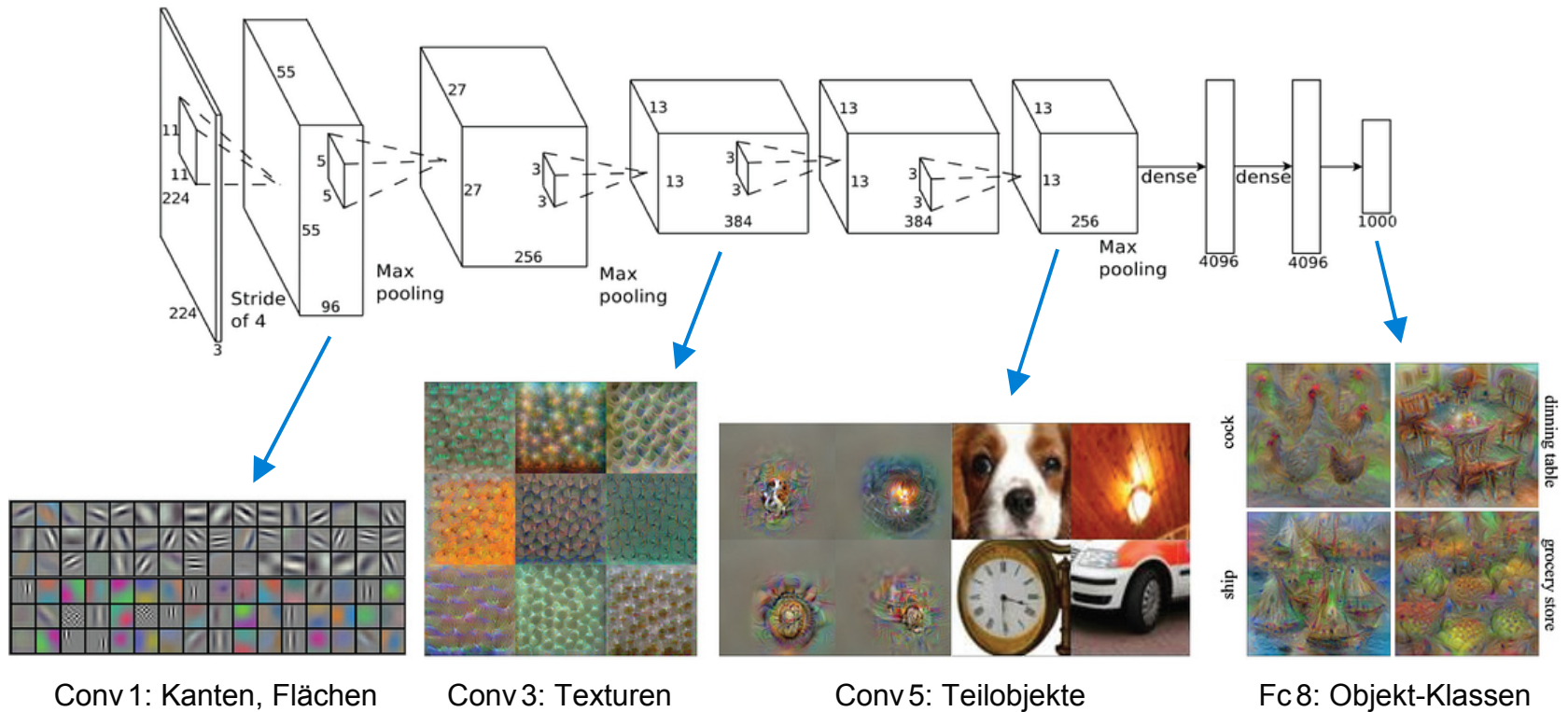


## Aufgabe

- Klassifikation
- 1,4 Mio. Bilder
- 1000 Klassen
- Top-5-Fehler
- Proband: ~5%



# Lernen von Merkmalen



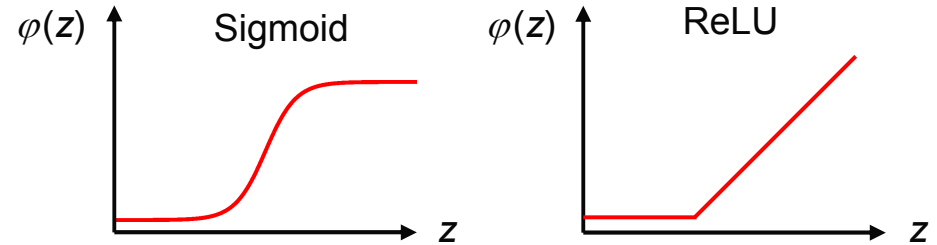
- Deep convolutional neural network (CNN)
- End-to-end learning

[http://vision03.csail.mit.edu/cnn\\_art/](http://vision03.csail.mit.edu/cnn_art/)

# Grund Nr. 1: Algorithmen

## Verschwindende Gradienten

- Sättigungsfreie Aktivierung
- Initialisierung, Normierung

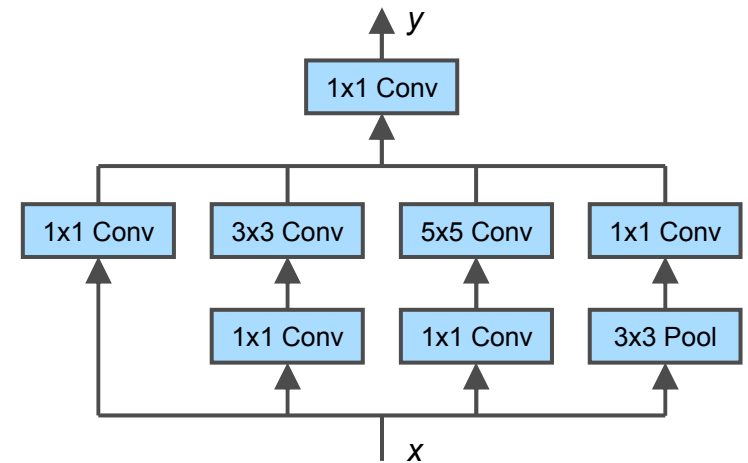


## Parameter-Anzahl

- Stochastisches Training, Mini-Batches
- Regularisierung: Dropout

## Netz-Architektur

- Faltung und Pooling
- Inception-Module, Residuen-Verbindungen



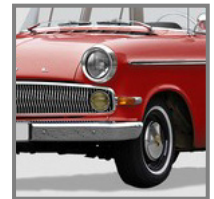
# Grund Nr. 2: Daten



<https://www.cityscapes-dataset.com/>

## Bereitstellen von Trainingsdaten

- Automatisiertes Sammeln
- Labeling (Objekte/Klassen markieren)
- Künstliches Variieren und Überlagern
- Entwicklungsprozess: Daten ↔ Algorithmen



<https://pixabay.com/>

# Grund Nr. 3: Hardware

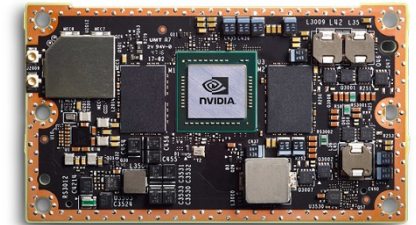
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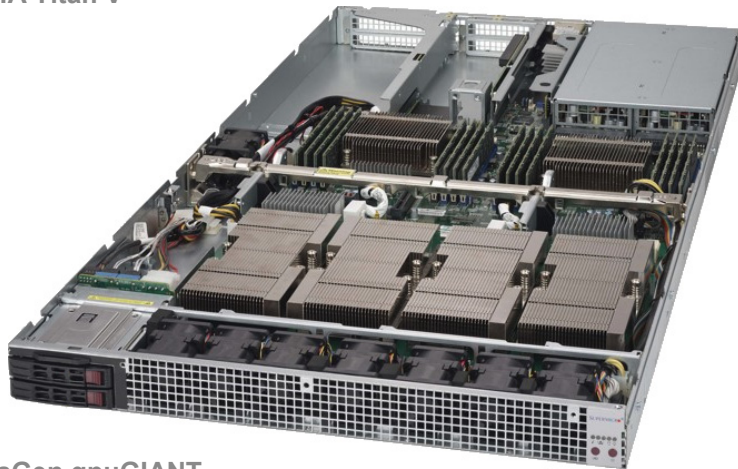
NVIDIA Titan V



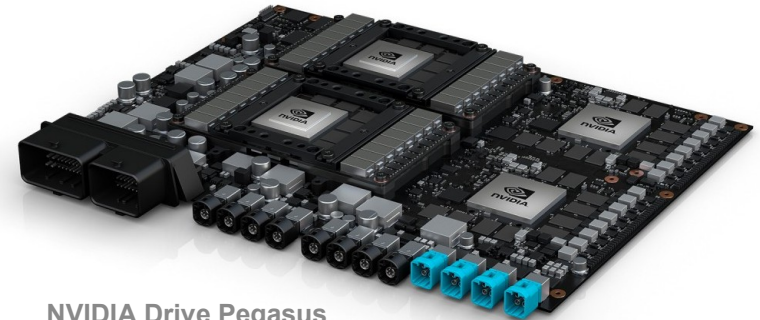
sysGen devCUBE



NVIDIA Jetson TX2

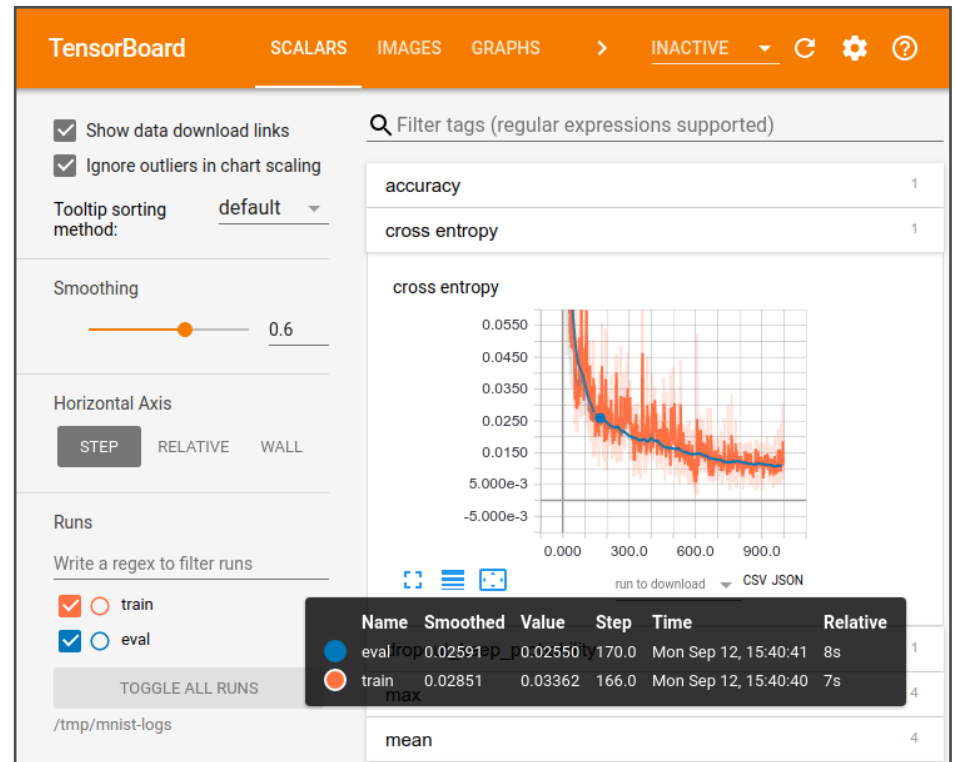


sysGen gpuGIANT



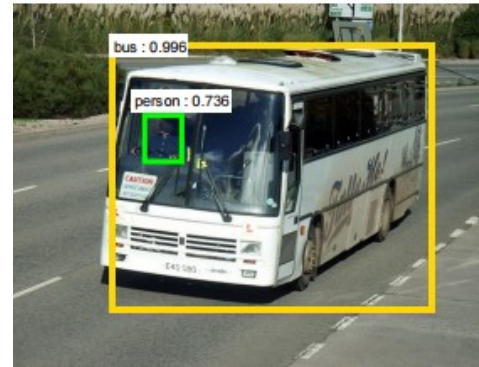
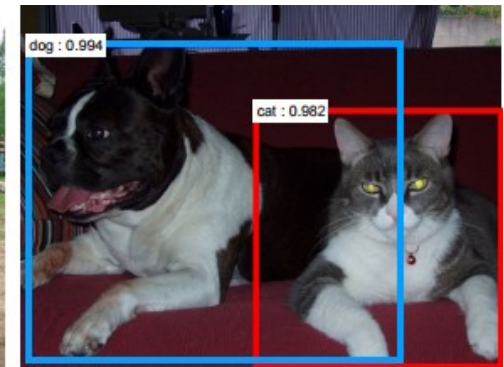
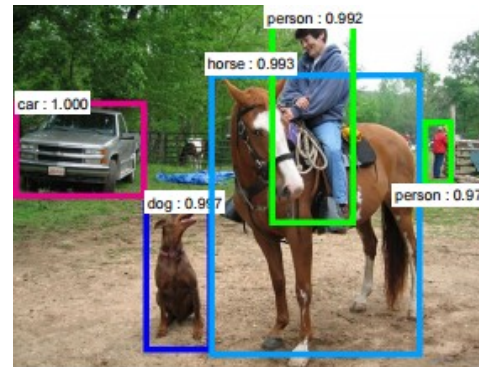
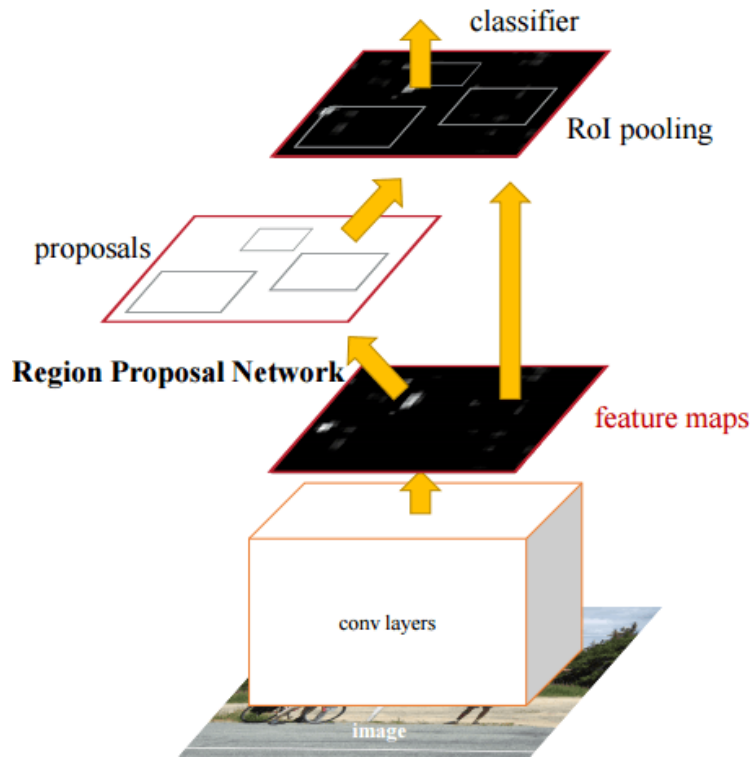
NVIDIA Drive Pegasus

# Grund Nr. 4: Software



<http://www.tensorflow.org/>

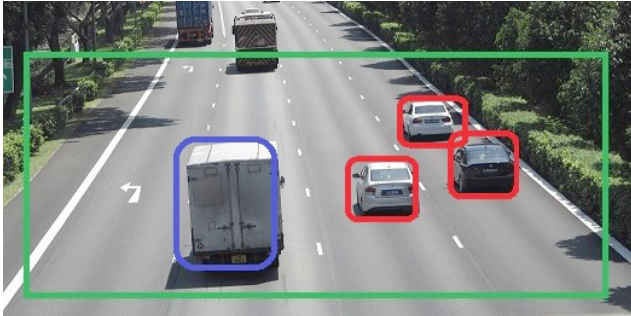
# Objekt-Detektion



Ren et al. (2016): Faster R-CNN

# Erfassen von Verkehrsszenen

## Fahrzeuge



Zhou et al. (2016)

## Fußgänger



Angelova et al. (2015)

## Straßen-Segmentierung



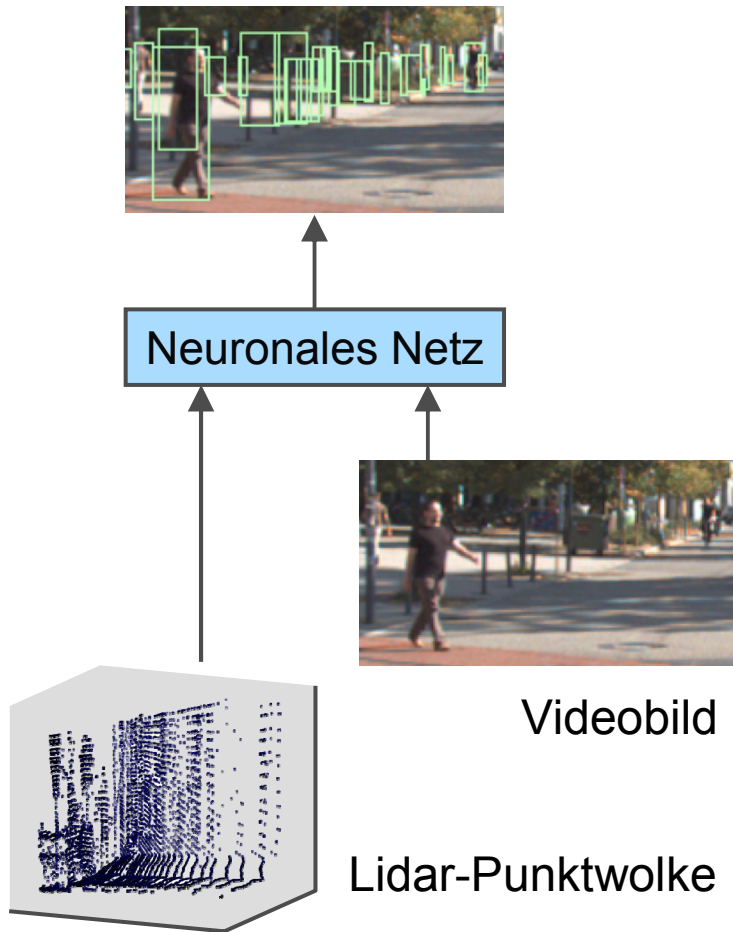
Wang et al. (2018)

## Verkehrszeichen

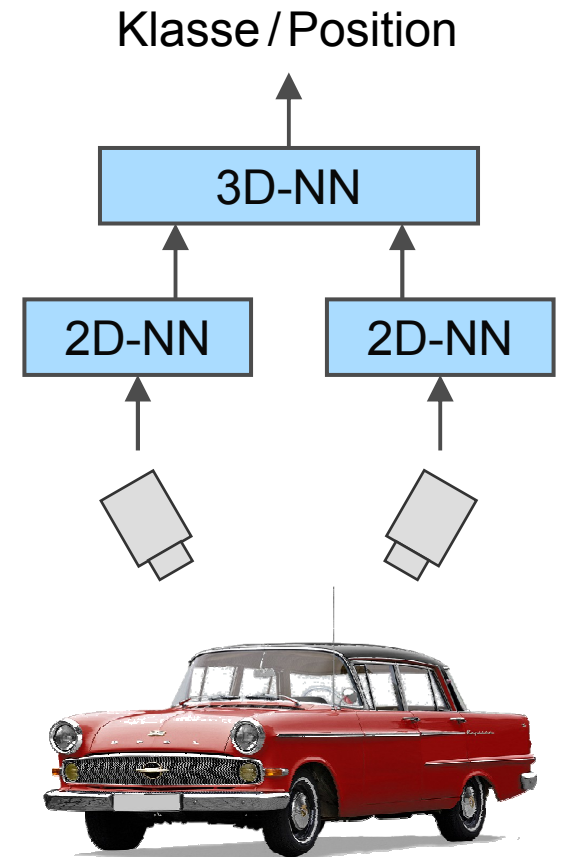


Zhu et al. (2016)

# Fusion mehrerer Datenquellen



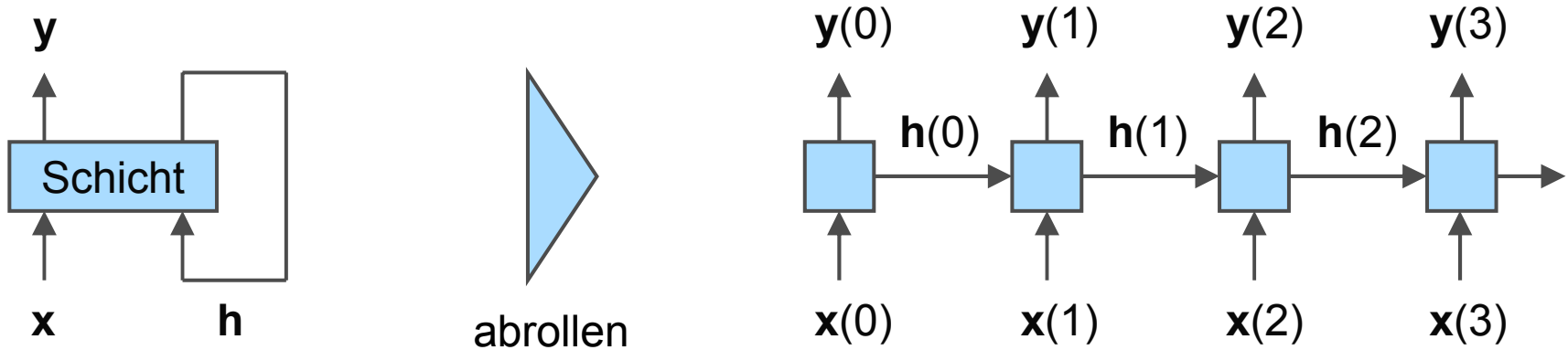
Matti et al. (2017)



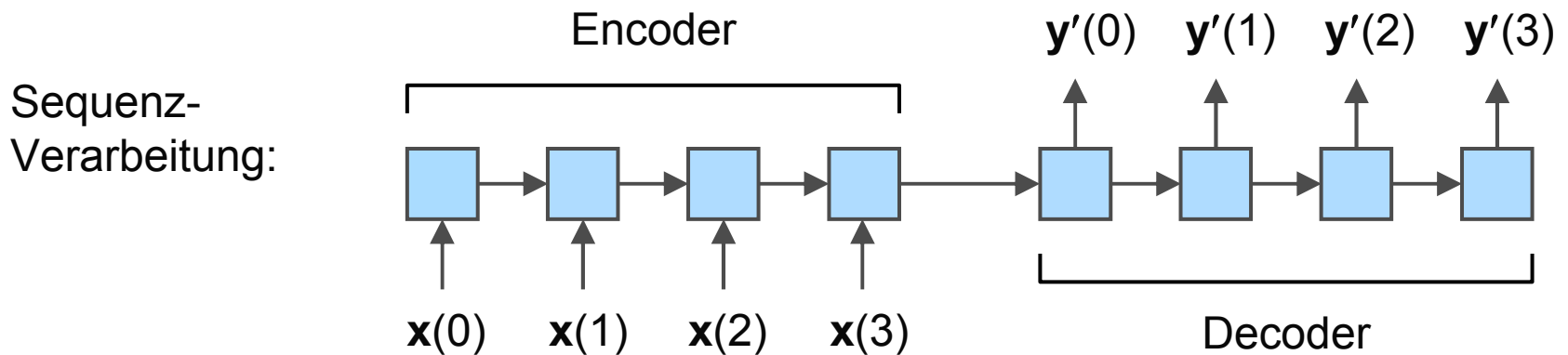
Multiview- / Stereokamera



# Rekurrente neuronale Netze (RNNs)



Training: „Backpropagation Through Time“ (BPTT)



# Bildbeschreibung (CNN-RNN)

Describes without errors



A person riding a motorcycle on a dirt road.

Describes with minor errors



Two dogs play in the grass.

Somewhat related to the image



A skateboarder does a trick on a ramp.

Unrelated to the image



A dog is jumping to catch a frisbee.



A group of young people playing a game of frisbee.



Two hockey players are fighting over the puck.



A little girl in a pink hat is blowing bubbles.



A refrigerator filled with lots of food and drinks.



A herd of elephants walking across a dry grass field.



A close up of a cat laying on a couch.



A red motorcycle parked on the side of the road.

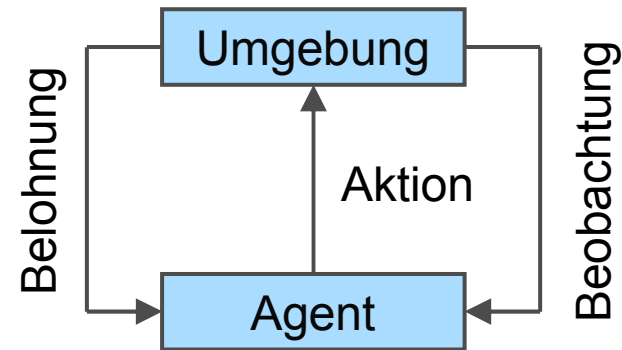


A yellow school bus parked in a parking lot.

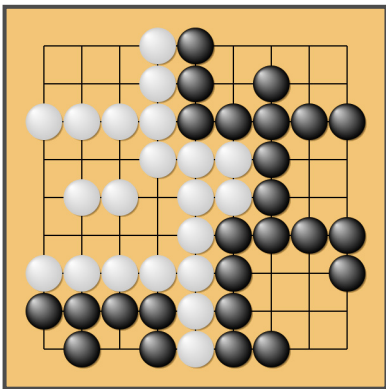
<http://research.googleblog.com/2014/11/>

# Reinforcement Learning

## Lernen durch Belohnung

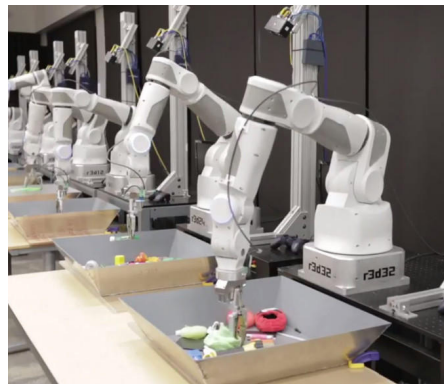


### Spielstrategien



<https://de.wikipedia.org/>

### Roboter-Bewegung



Levine et al. (2017)

### Autonomes Fahren



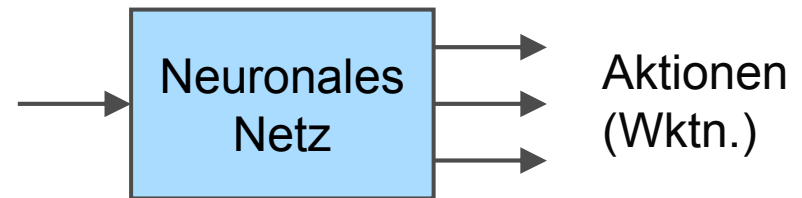
<https://waymo.com/>

# Deep Reinforcement Learning

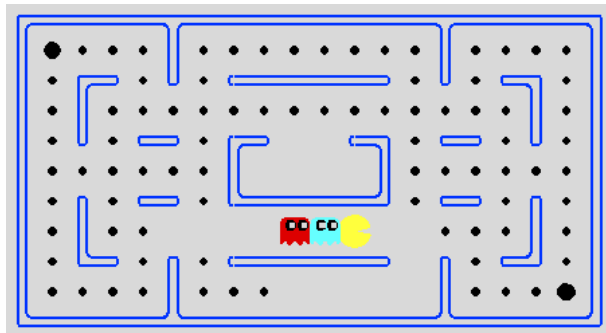
## Neuronale Netze als Policy



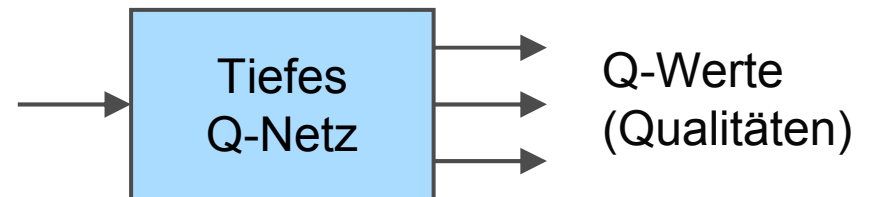
<https://devblogs.nvidia.com/>



## Deep Q-Learning



<https://www.cs.washington.edu/>



# Zusammenfassung

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## Deep Learning

- Neuronale Netze mit vielen Schichten
- Lernen aller Merkmale „end-to-end“
- Gründe: Algorithmen, Daten, Hardware, Software

## Anwendungen

- Bilder/Videos: Objekt- und Gesichtserkennung, Segmentierung, Retusche, Superresolution, ...
- Sequenzen: Spracherkennung, Übersetzung, Dialog
- Reinforcement Learning: Spiele, Robotik, Regelung

## Konsequenzen

- Mehr Automatisierung bei besserer Qualität
- Verschiebung von Algorithmen zu Daten