Künstliche Intelligenz im Browser?!

Wie auch du diese Technologie ohne tiefes Verständnis nutzen kannst.

Max Rose

B. Sc. Angewandte Informatik

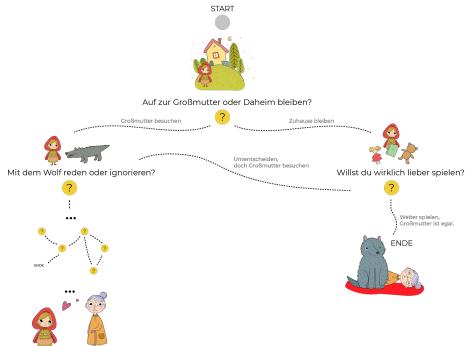
Schwerpunkt: Virtuelle Realität

Interesse: Künstliche Intelligenz

Gründer von audory

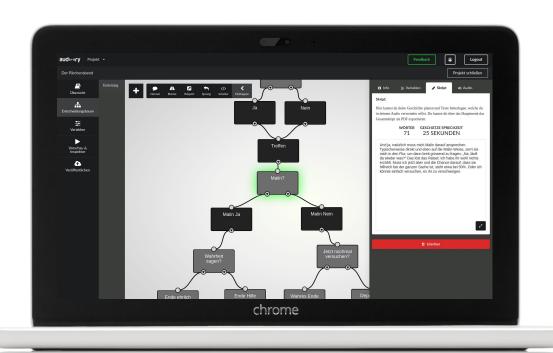


audory - Deine Plattform für interaktive Hörspiele





audory - Deine Plattform für interaktive Hörspiele







Ziele des Talks

1. Grundlagen verstehen

2. Startpunkt bekommen

3. Motivation zur Anwendung

Gliederung

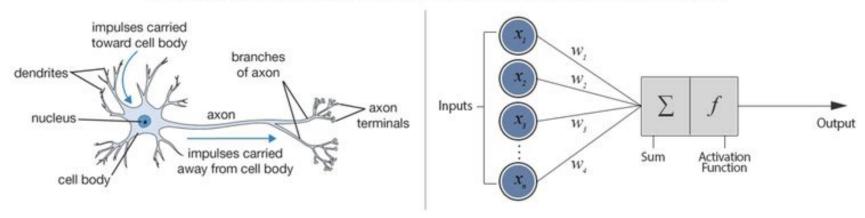
- 1. Grundlagen Neuronale Netze
 - a. Aufbau/Funktion eines Neurons
 - b. Architektur von Neuronalen Netzen
- 2. Training von Neuronalen Netzen
 - a. Daten sammeln & aufbereiten
 - b. Architektur designen
 - c. Model trainieren/testen

- 3. Umsetzung mit Tensorflow.js
 - a. Vorstellung Tensorflow.js
 - b. Teachable Machine Live Demo
- 4. Ausblick
 - a. Weitere Projektideen
- 5. Diskussion

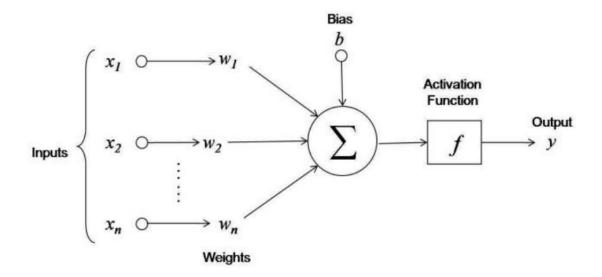
Grundlagen Neuronale Netze

Aufbau/Funktion eines Neurons

Biological Neuron versus Artificial Neural Network



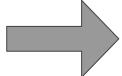
Aufbau/Funktion eines Neurons



$$y = f(\langle \mathbf{w}, \mathbf{x}
angle + b) = f(\sum_{i=1}^N w_i \, x_i + b)$$

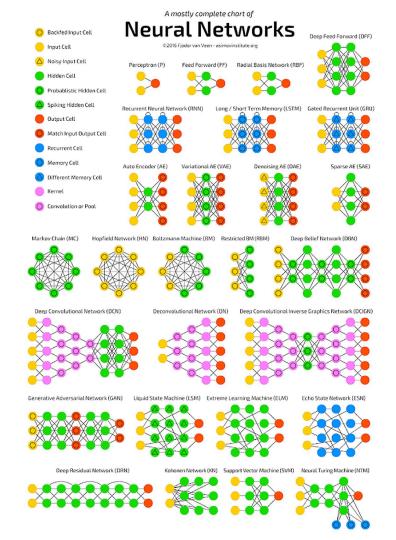


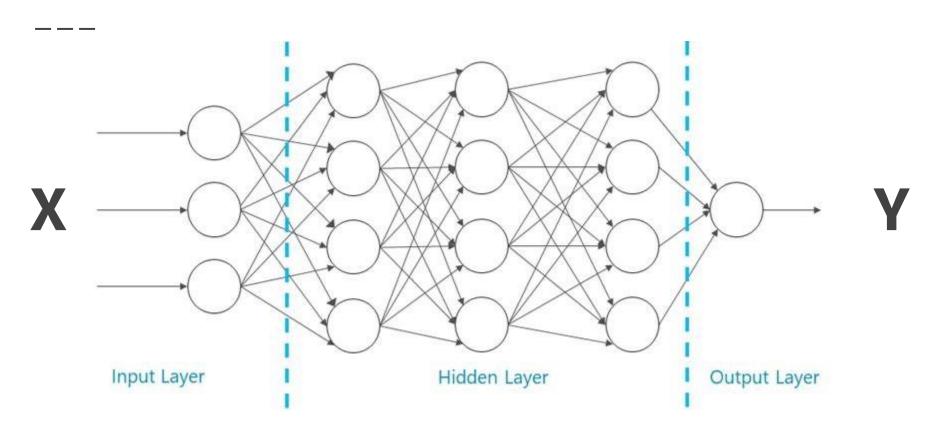






Output Hidden Input layer layer layer X

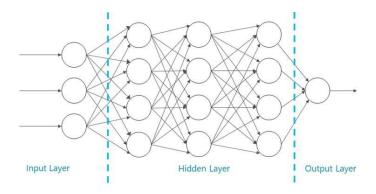




Training von Neuronalen Netzen

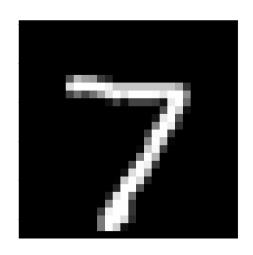
Beispiel

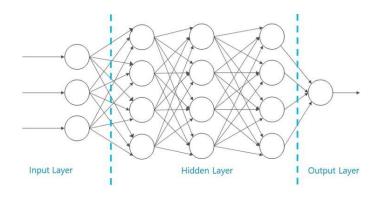




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Beispiel



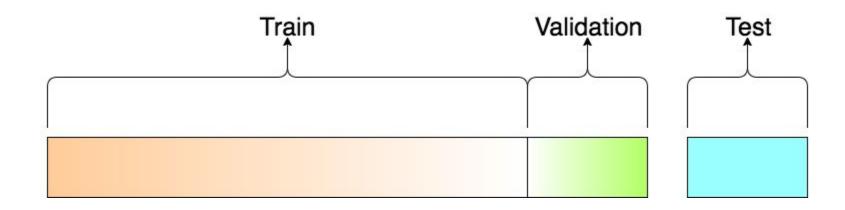


7

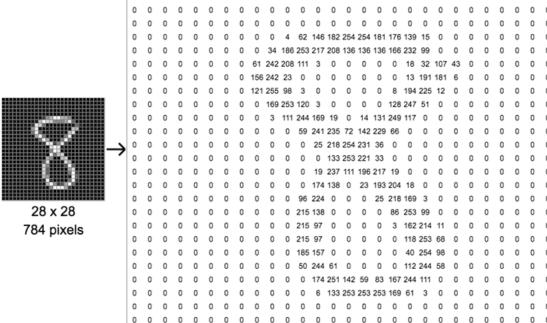
Daten sammeln & aufbereiten

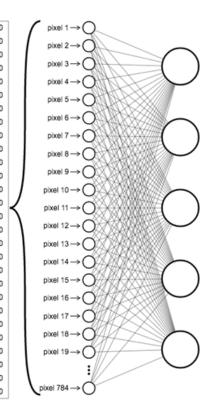
Output (Label) Input 0000000000 11/1/////// 222222222 23333333333 55555555555 6666666666 7 7 7 7 **7 7** 7 7 7 7 7 7 8888888888 999999999

Daten sammeln & aufbereiten



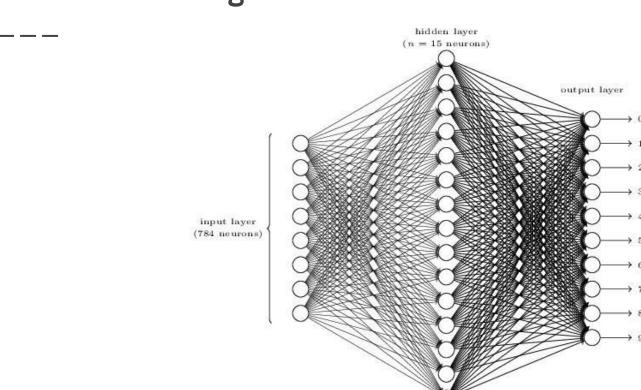
Daten sammeln & aufbereiten







Architektur designen



Architektur designen

- 1. Umso komplexer das Problem -> Umso mehr Neuronen nötig
- 2. Umso mehr Neuronen -> Umso komplexer das Model
- 3. Umso komplexer das Model -> Umso mehr Trainingsdaten

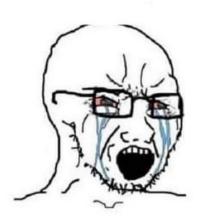
Model trainieren/testen

model.fit(X, Y);

Model trainieren/testen

```
Y = model.predict(X);
```

So you just
stack some
stupid layers,
train on some
GBs of data and
it works better
than handcrafted
techniques?



YES



Umsetzung mit Tensorflow.js

Vorstellung Tensorflow.js



https://www.tensorflow.org/js/

yarn add @tensorflow/tfjs



Vorstellung Tensorflow.js



```
const tf = require('@tensorflow/tfjs');
// Train a simple model:
const model = tf.sequential();
model.add(tf.layers.dense({units: 15, activation: 'relu', inputShape: [784]}));
model.add(tf.layers.dense({units: 10, activation: 'sigmoid'}));
model.compile({optimizer: 'sgd', loss: 'meanSquaredError'});
const xs = tf.randomNormal([100, 784]);
const ys = tf.randomNormal([100, 10]);
model.fit(xs, ys, {
 epochs: 100,
 callbacks: {
   onEpochEnd: (epoch, log) => console.log(`Epoch ${epoch}: loss = ${log.loss}`)
}
});
```

Teachable Machine



DEMO

Ausblick

Weitere Projektideen

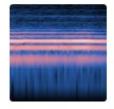


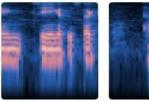




Image Project

Teach based on images, from files or your webcam.







Audio Project

Teach based on one-second-long sounds, from files or your microphone.







Pose Project

Teach based on images, from files or your webcam.

Weitere Projektideen

https://thecodingtrain.com/

ML5 BEGINNERS GUIDE

Beginners Guide to Machine Learning in JavaScript



A BEGINNER'S GUIDE TO MACHINE LEARNING WITH ML5.JS

31 Jul 2018

Welcome to "A Beginner's Guide to Machine Learning in JavaScript"! In this series, I'll teach the concepts behind machine learning using the ml5.js library. #machinelearning #javascript #ml5 #p5js.



ML5.JS: IMAGE CLASSIFICATION WITH MOBILENET

01 Aug 2018

In this video, I use the "pre-trained" MobileNet model to classify the content of an image. #machinelearning #mobilenet #imageclassification #ml5 #p5js.



ML5.JS: WEBCAM IMAGE CLASSIFICATION

02 Aug 2018

In this video, I discuss image classification with MobileNet using real-time video. #machinelearning #mobilenet #imageclassification #mI5 #p5js #webcam.

Diskussion



This is why we train autonomous cars in San Francisco

Traducir Tweet













max.rose xamesor







Max Rose