



Practical Machine Learning

Introduction



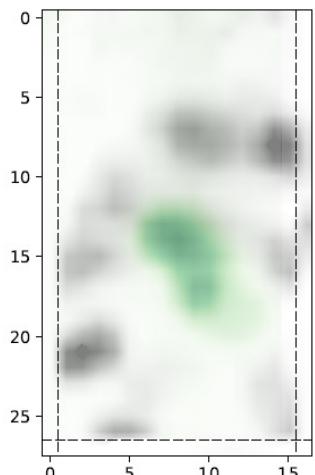
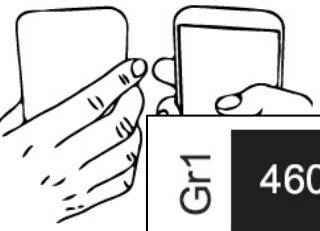
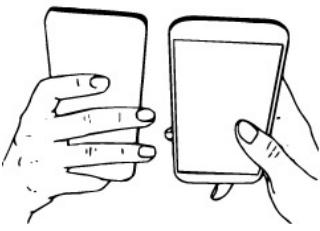
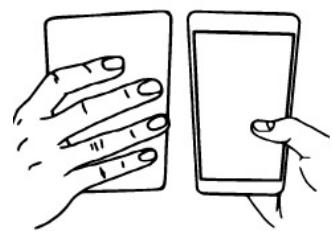
Sven Mayer

Where can we find ML used?

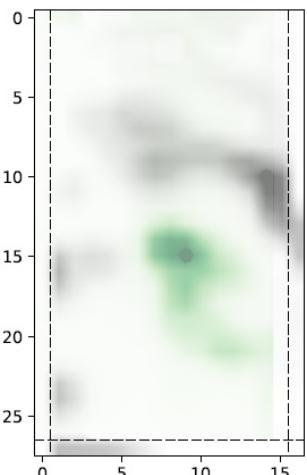
- Self driving cars (Tesla)
- Voice interfaces (Alexa, Siri)
- Face recognition (Google Photos)
- Recomender systems (Netflix, Amazon)
- Games (AlphaGo)
- Character recognition (Post offices)
- Banking systems
- Medical diagnosis
- ML for Human-Computer Interaction  **Focus**

Introduction into Machine Learning

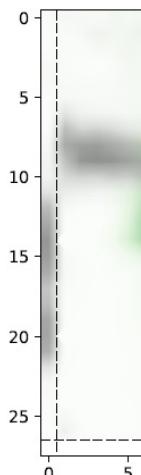
Examples



(a) Grip 1



(b) Grip 2



(c)

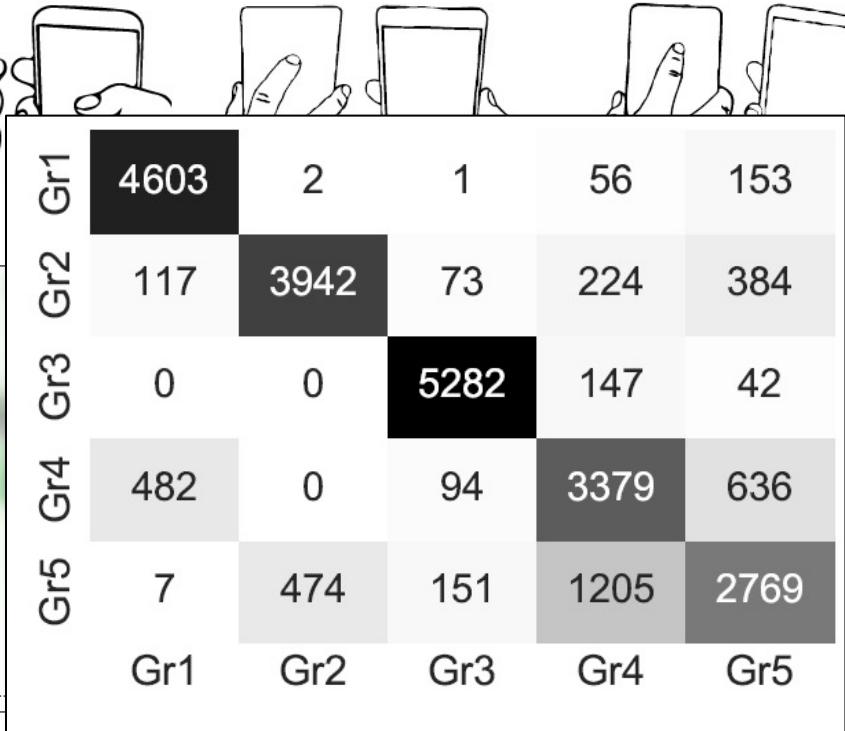


Figure 7: Confusion matrix showing the classification results for the exemplary grip classifier.

Huy Viet Le, Sven Mayer, Patrick Bader, and Niels Henning. 2017. A Deep Learning Model for Multi-Grip Classification on the Whole Device Surface. In Proceedings of MobileHCI '17, 10–14 September 2017, Berlin, Germany. ACM, New York, NY, USA, 1–11. DOI: <https://doi.org/10.1145/3078507.3078511>

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Examples



Huy Viet Le, Sven Mayer, Maximilian Weiß, Jonas Vogelsang, Henrike Weingärtner, and Niels Henze. 2020. Shortcut Gestures for Mobile Text Editing on Fully Touch Sensitive Smartphones. ACM ToCHI. DOI: <https://doi.org/10.1145/3396233>

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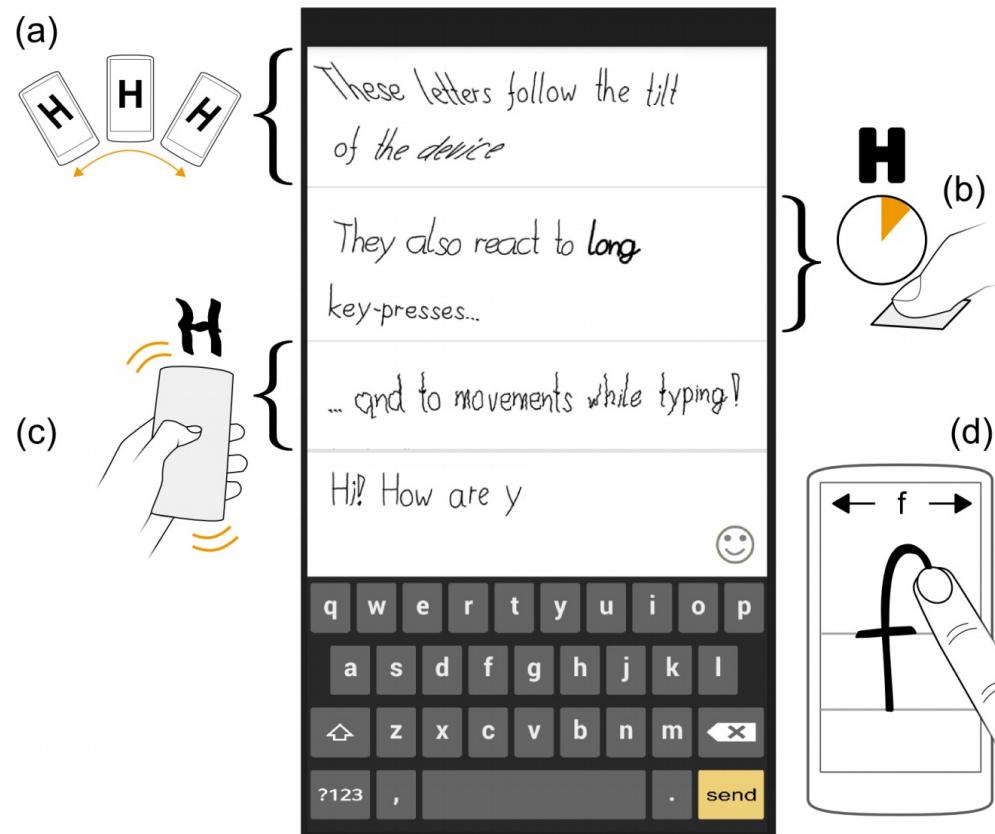
Examples



Huy Viet Le, Thomas Kosch, Patrick Bader, Sven Mayer, and Niels Henze. 2018. PalmTouch: Using the Palm as an Additional Input Modality on Commodity Smartphones. In Proceedings of CHI '18. Association for Computing Machinery, New York, NY, USA, Paper 360, 1–13. DOI: <https://doi.org/10.1145/3173574.3173934>

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Adaptive Keyboards



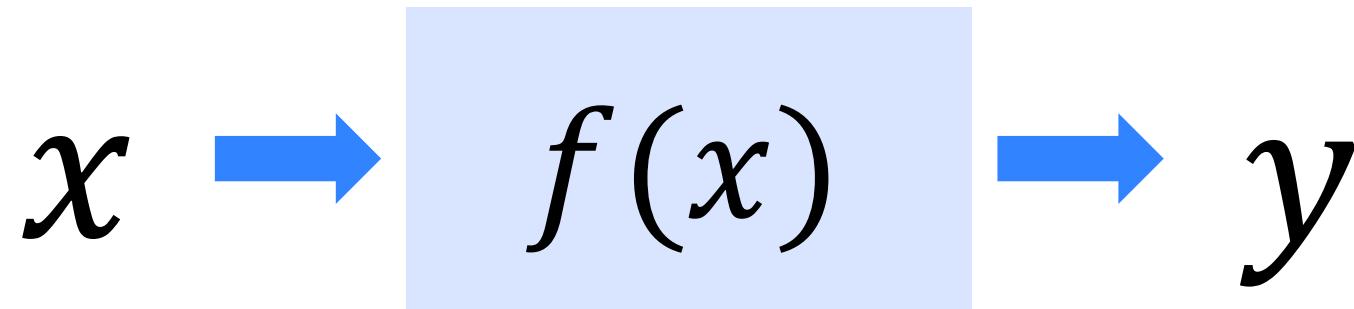
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Examples

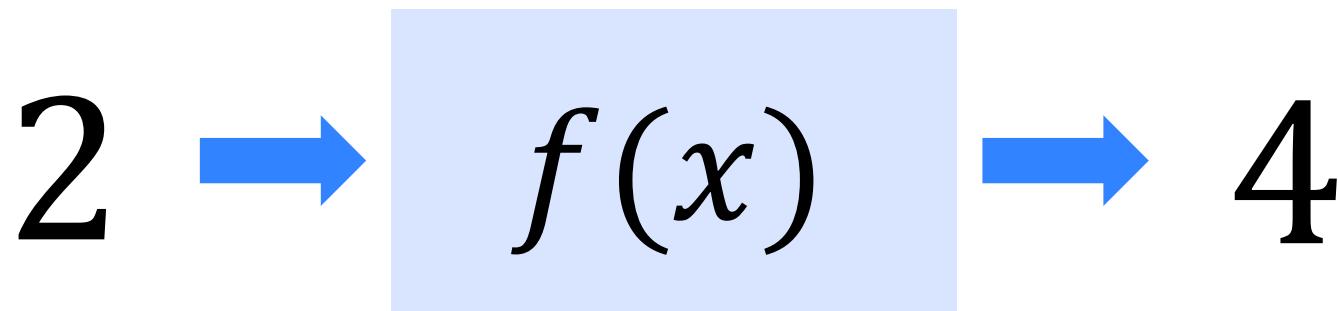


Xucong Zhang, Yusuke Sugano, and Andreas Bulling. 2019. Evaluation of Appearance-Based Methods and Implications for Gaze-Based Applications. In Proc. Of CHI '19. ACM. DOI: <https://doi.org/10.1145/3290605.3300646>

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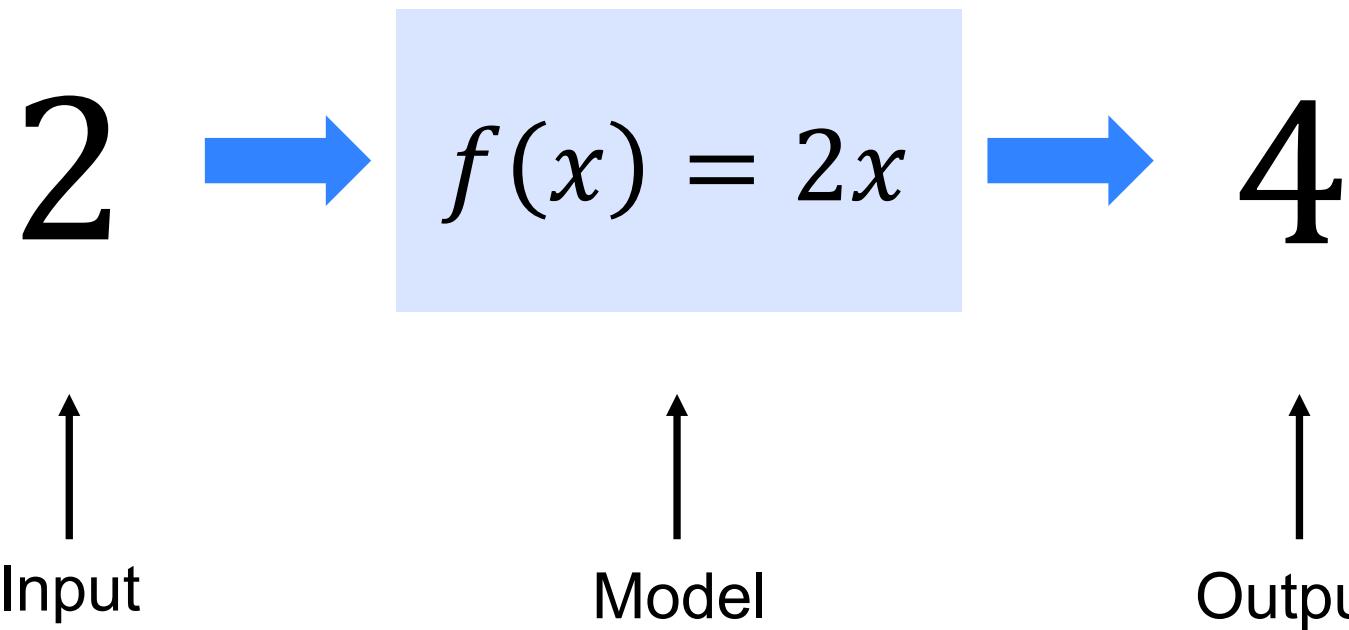
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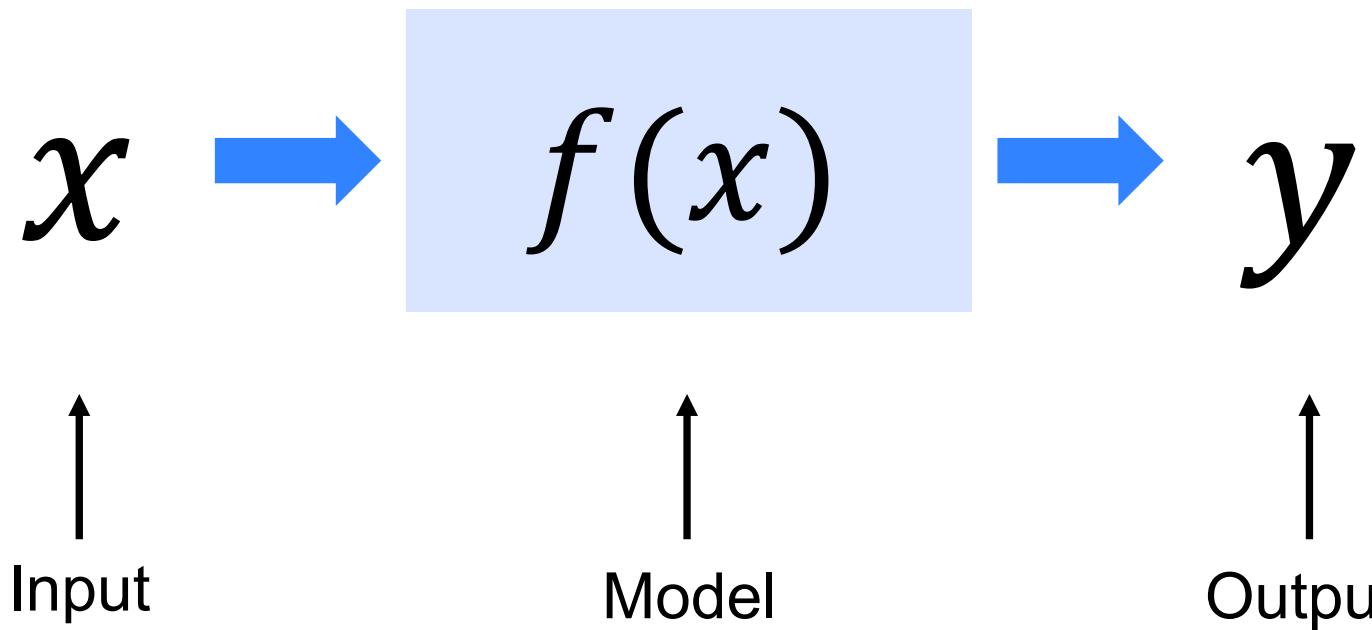
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$$2 \rightarrow f(x) = 2x \rightarrow 4$$

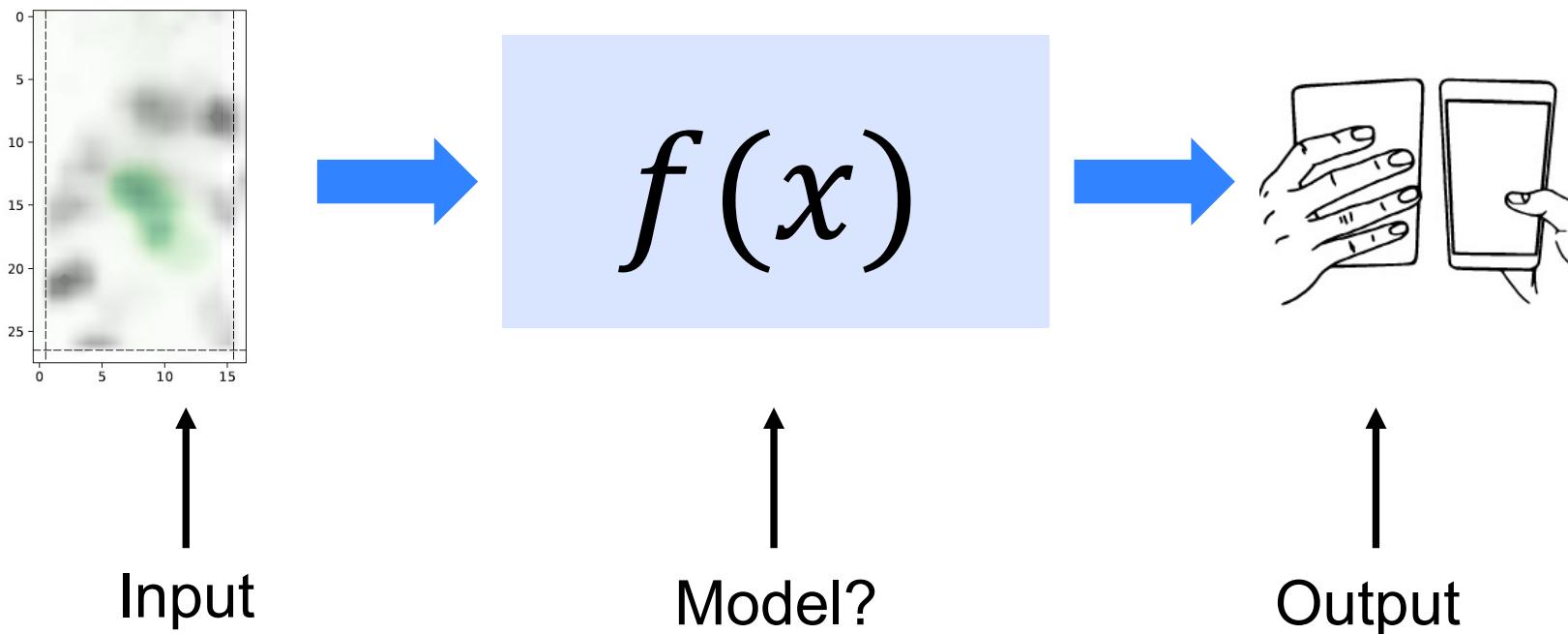
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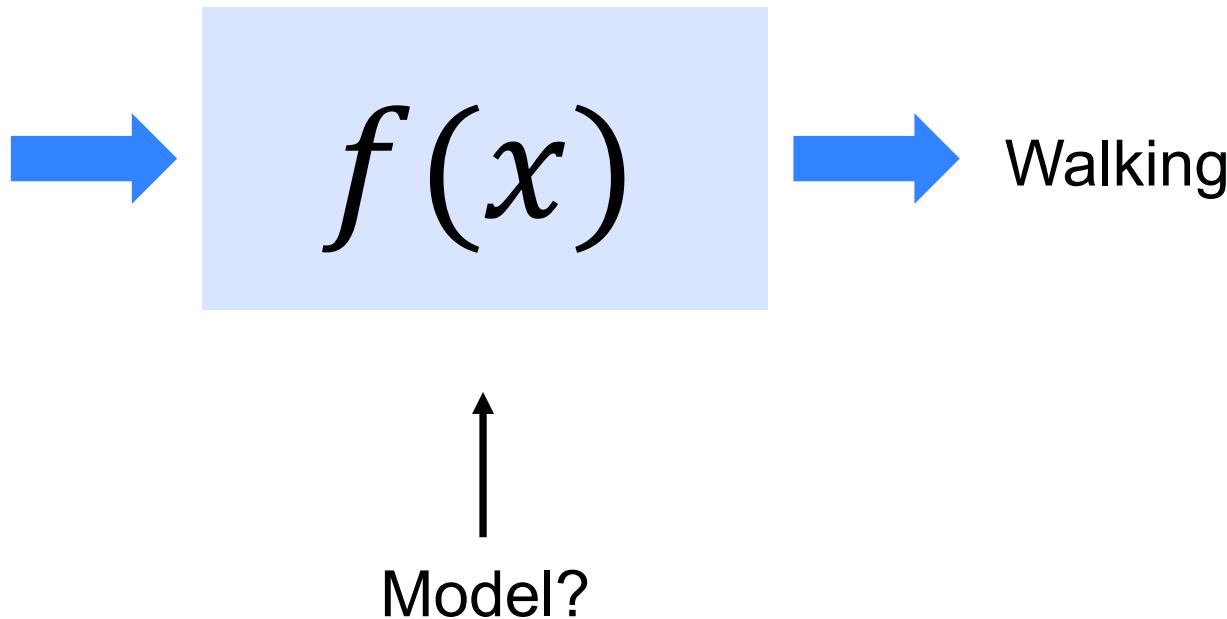
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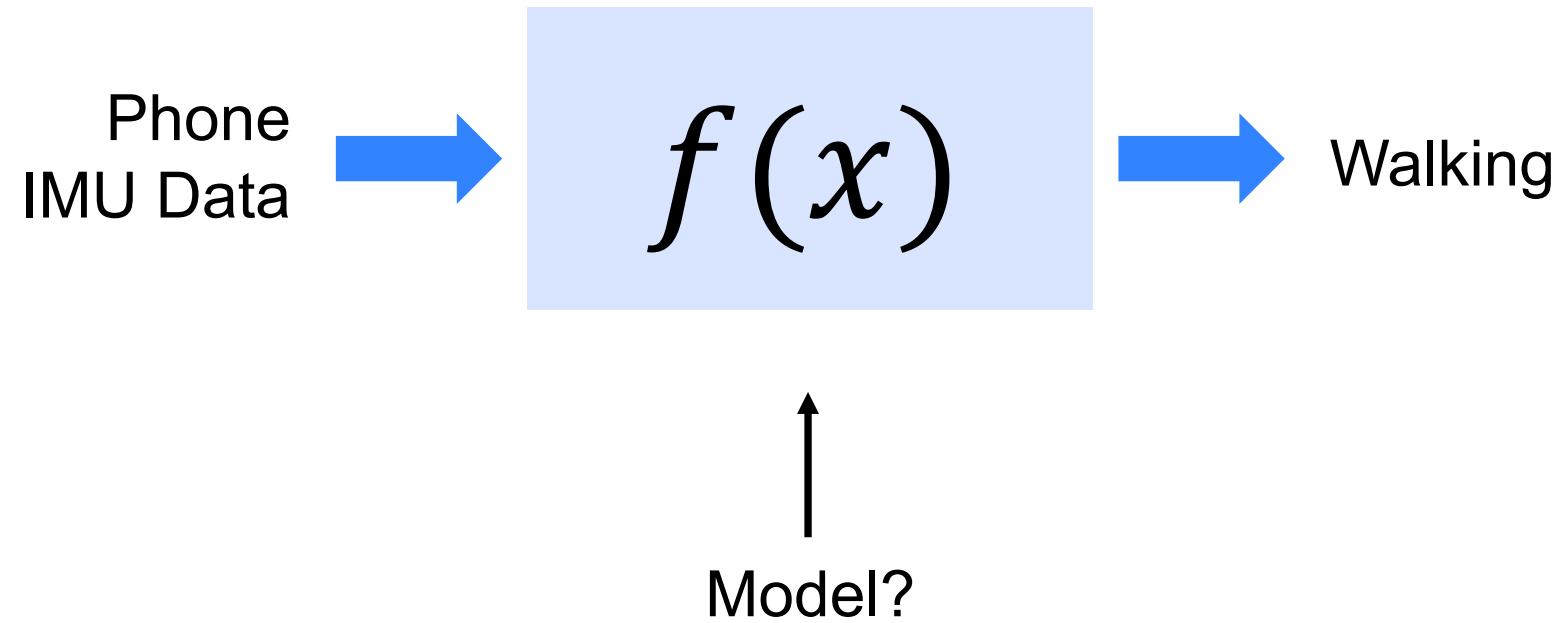


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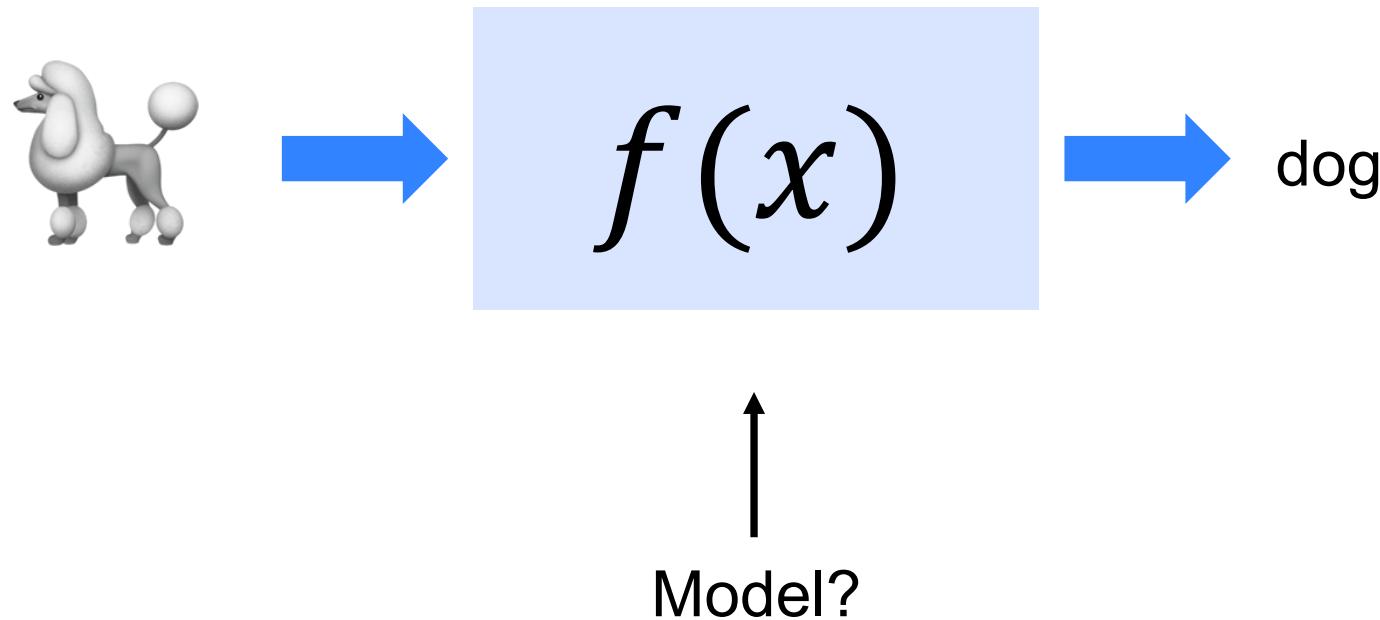
*human activity recognition

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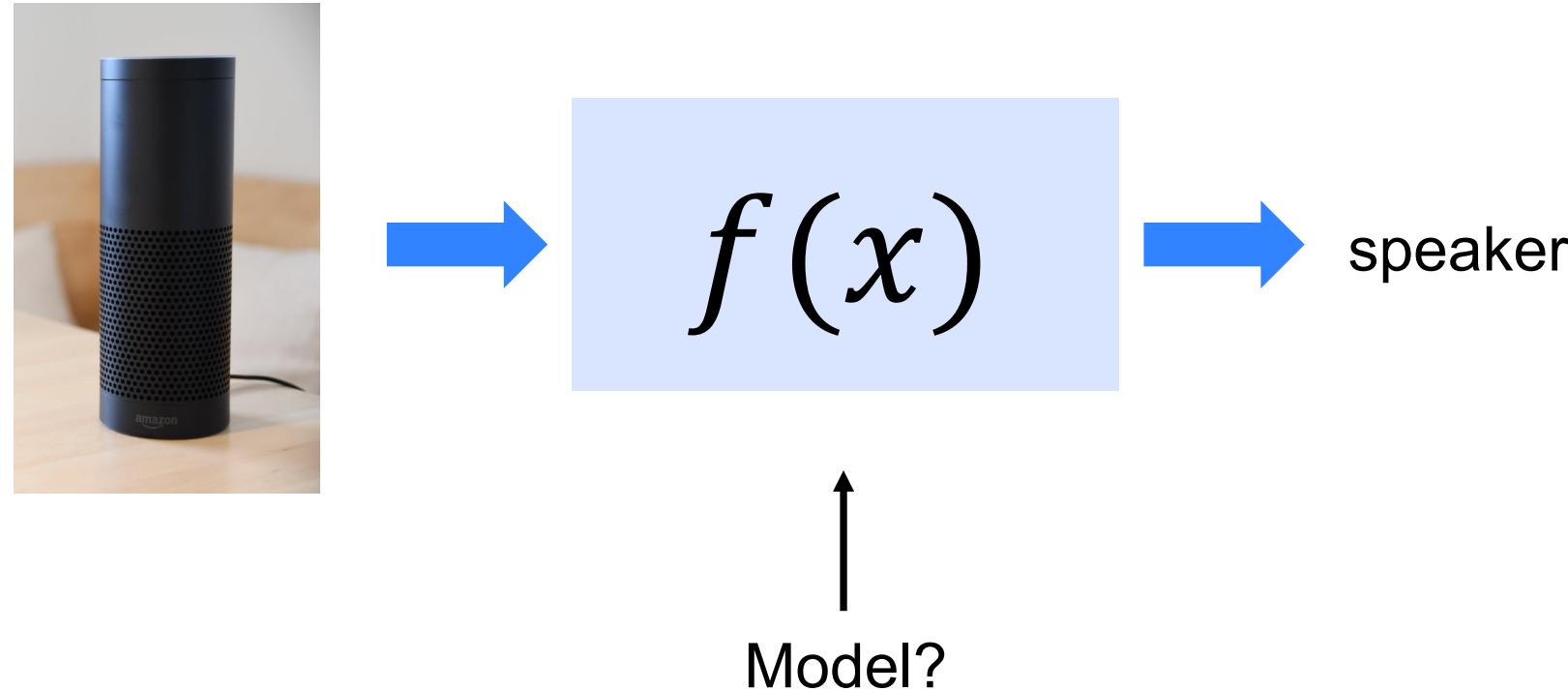
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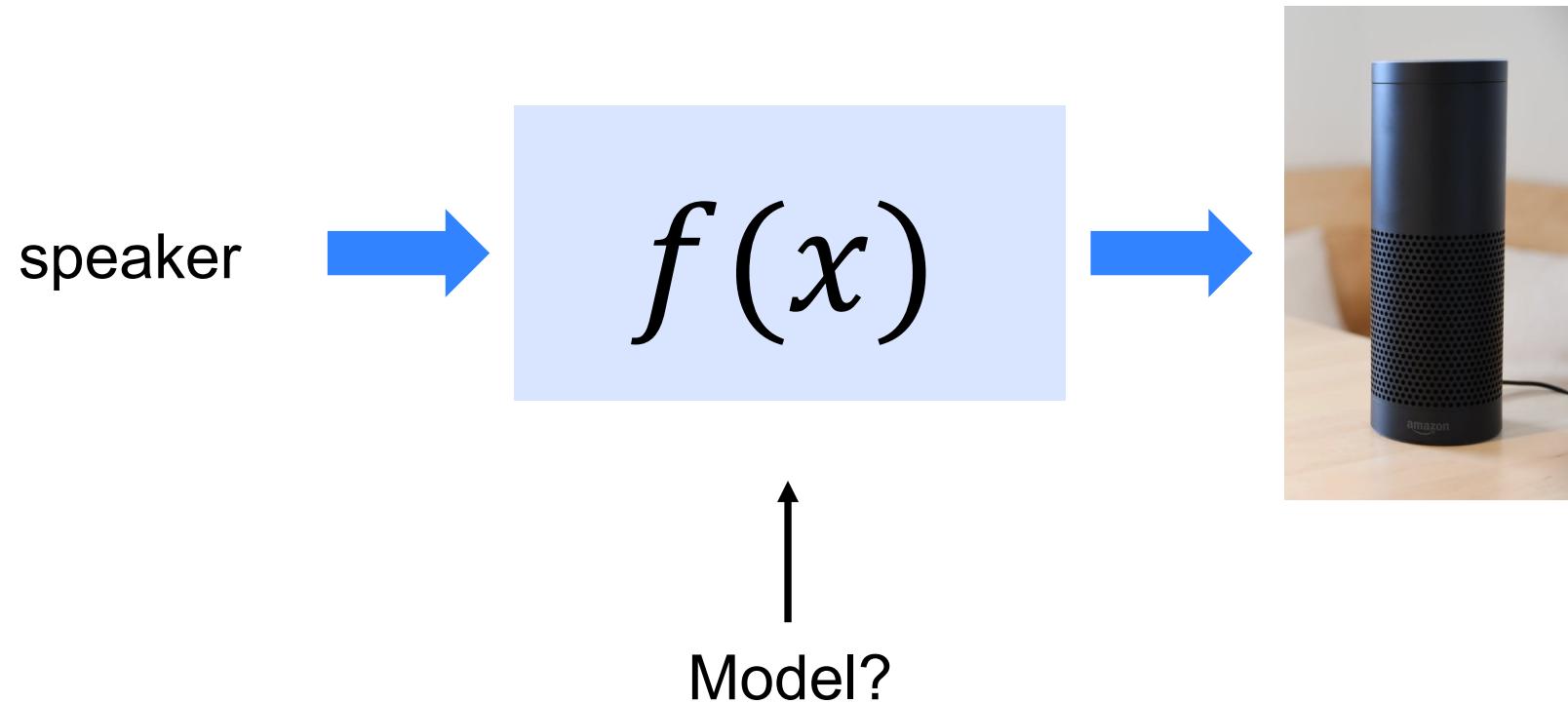
*object recognition

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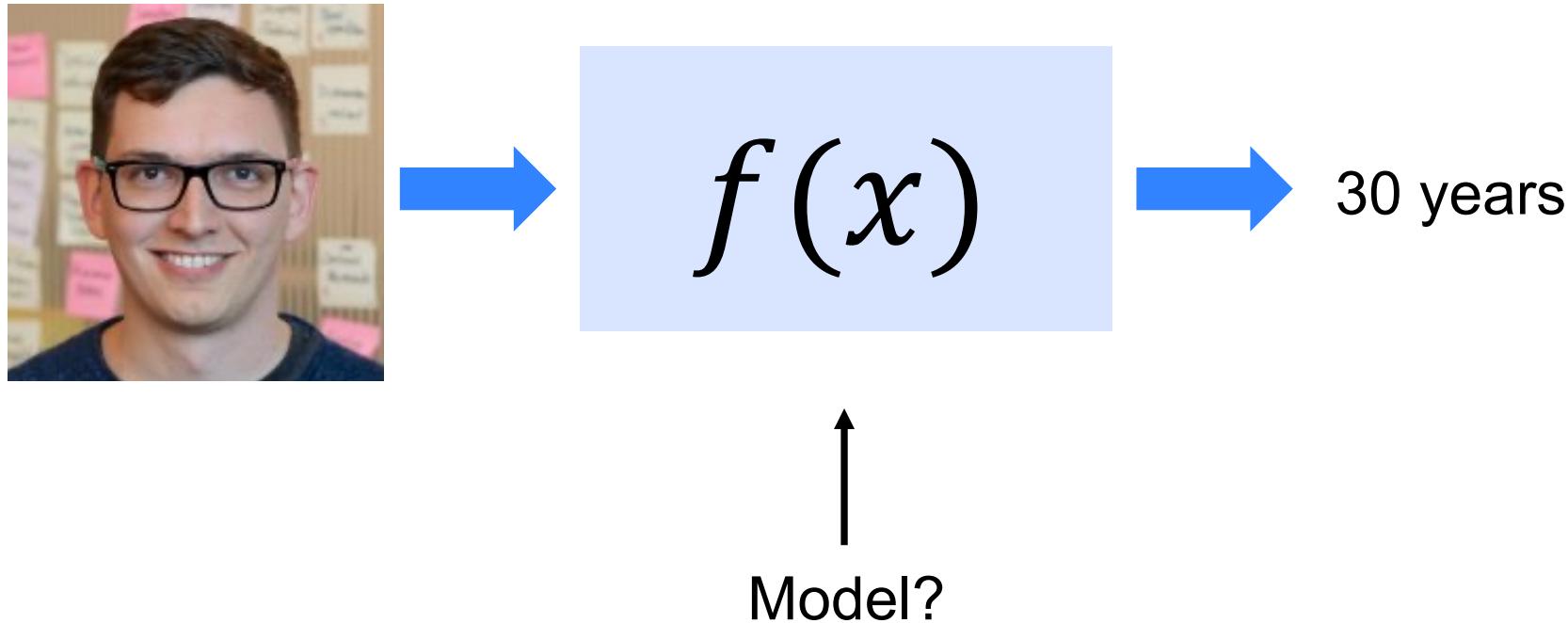
*object recognition

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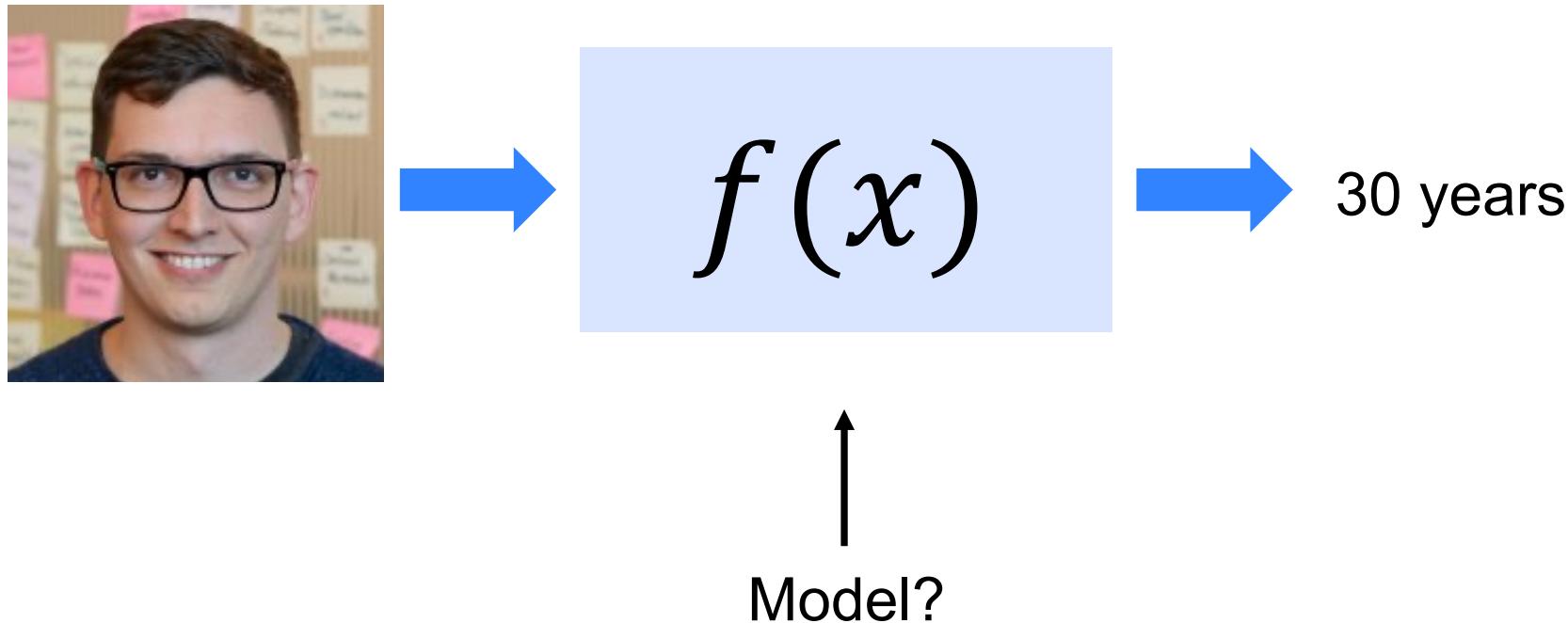
*Generative Adversarial Network (GAN)

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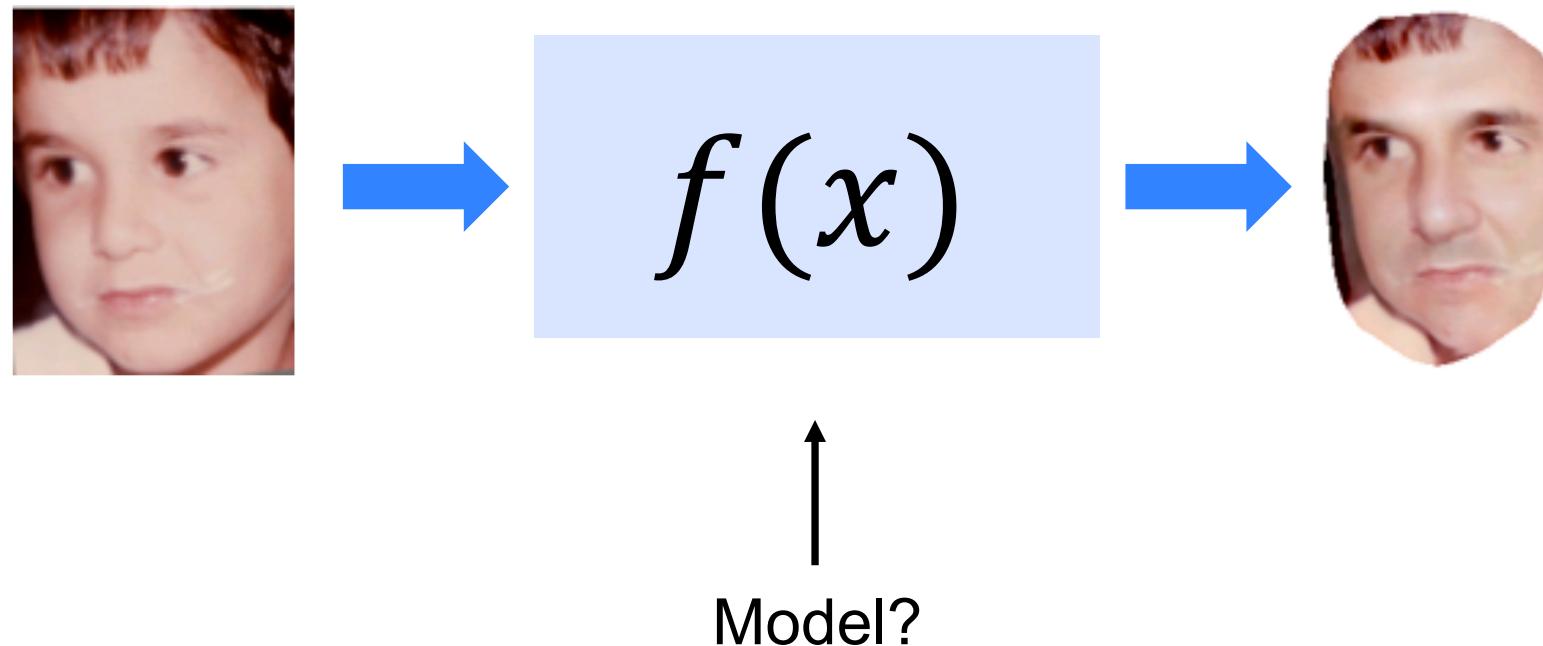
Gil Levi, and Tal Hassner. "Age and gender classification using convolutional neural networks." In Proceedings of the CVPR workshops. 2015. IEEE: <https://doi.org/10.1109/CVPRW.2015.7301352>

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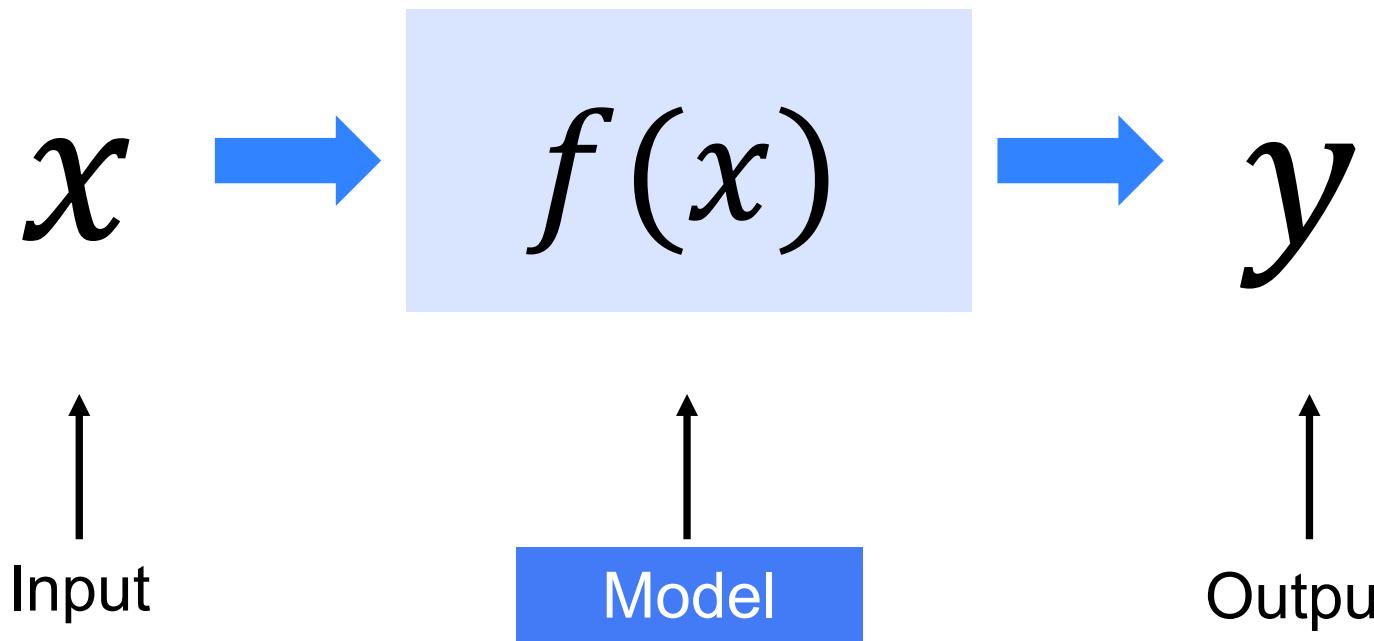
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Ira Kemelmacher-Shlizerman, Supasorn Suwajanakorn, and Steven M. Seitz. 2014. Illumination-aware age progression. In Proceedings of the CVPR 2014. IEEE. DOI: <https://doi.org/10.1109/CVPR.2014.426>

Introduction into Machine Learning



Models

- “Traditional” Machine Learning
 - Support Vector Machines
 - Decision Trees
 - Random Forest
 - ...
- “Deep” Learning Methods
 - Neuronal Networks
 - Convolutional Neuronal Networks
 - Recurrent Neural Network (RNN)
 - Generative Adversarial Network (GAN)
 - ...

Models

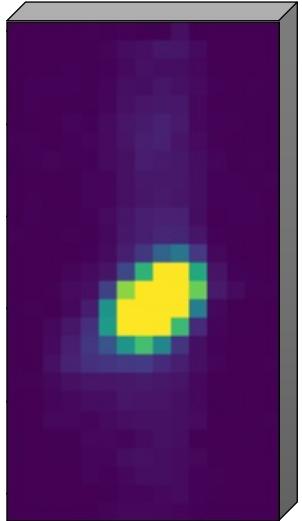
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We will look at some of them in this lecture.

We will focus mainly on Deep Learning Methods.

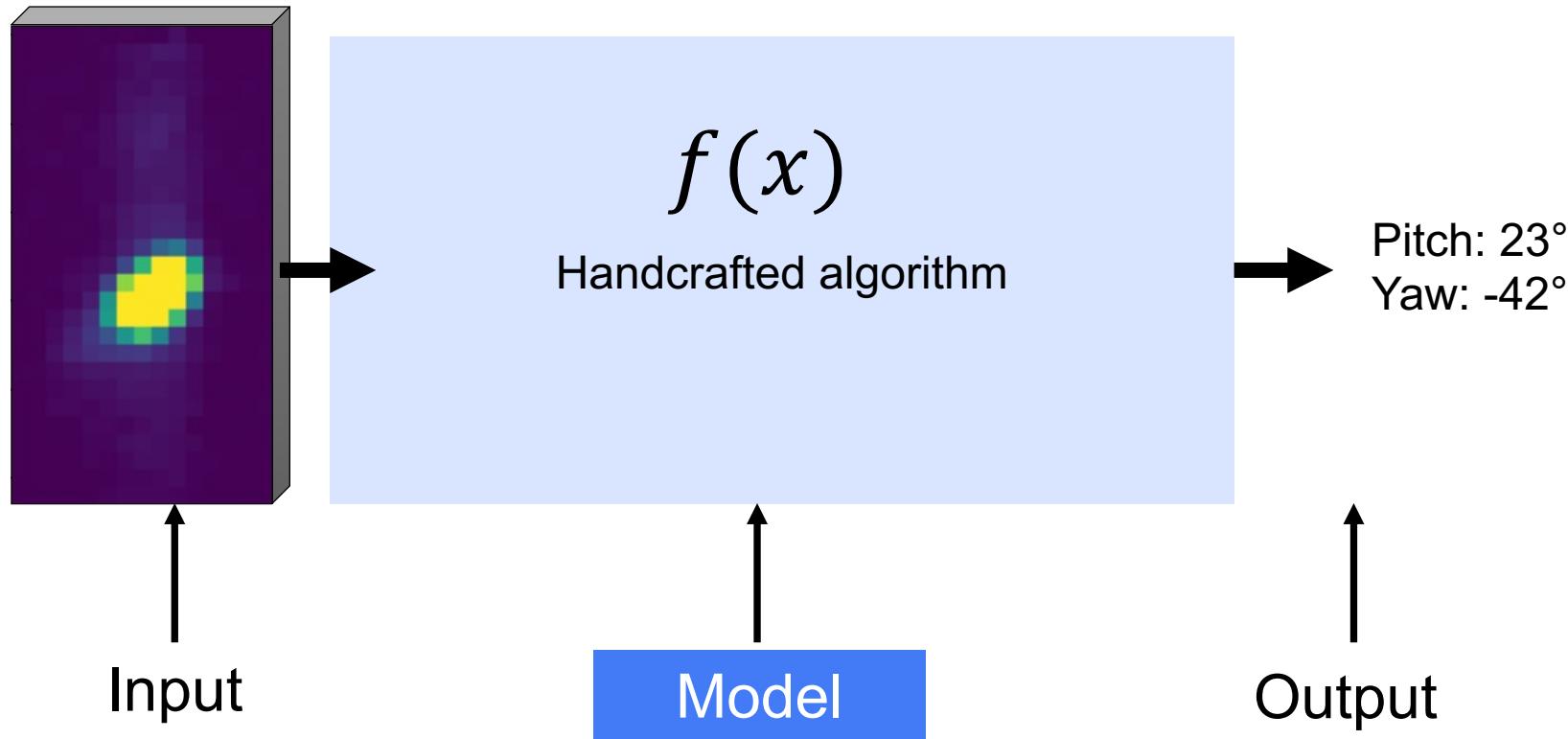
Example

Human-Computer Interaction



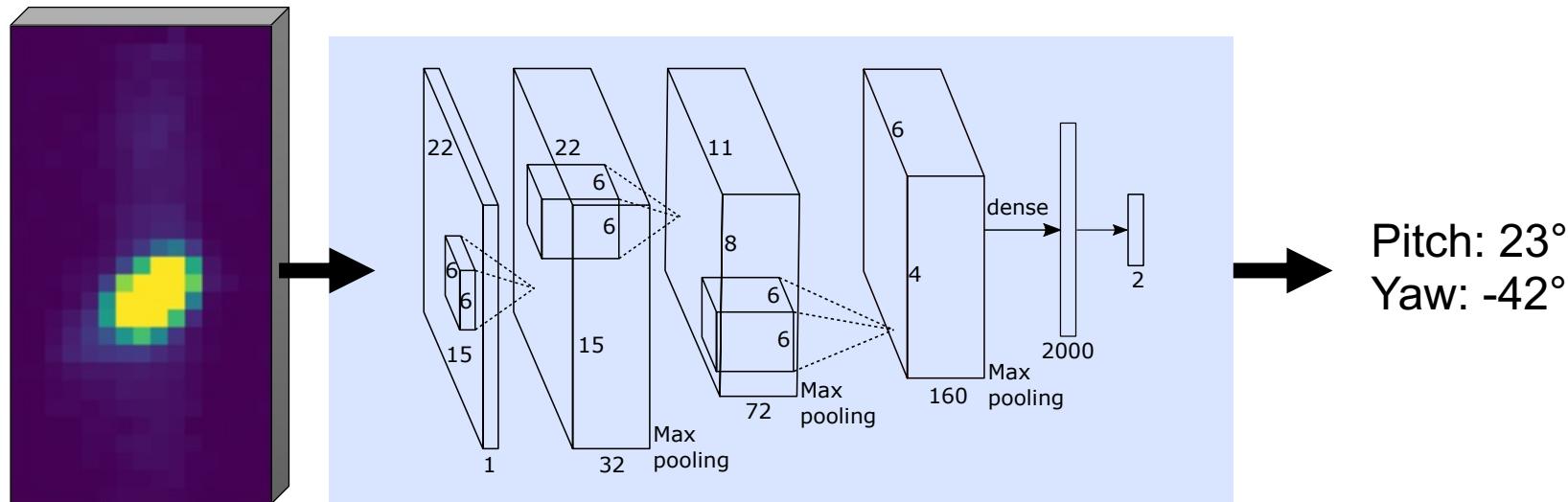
Example

Human-Computer Interaction



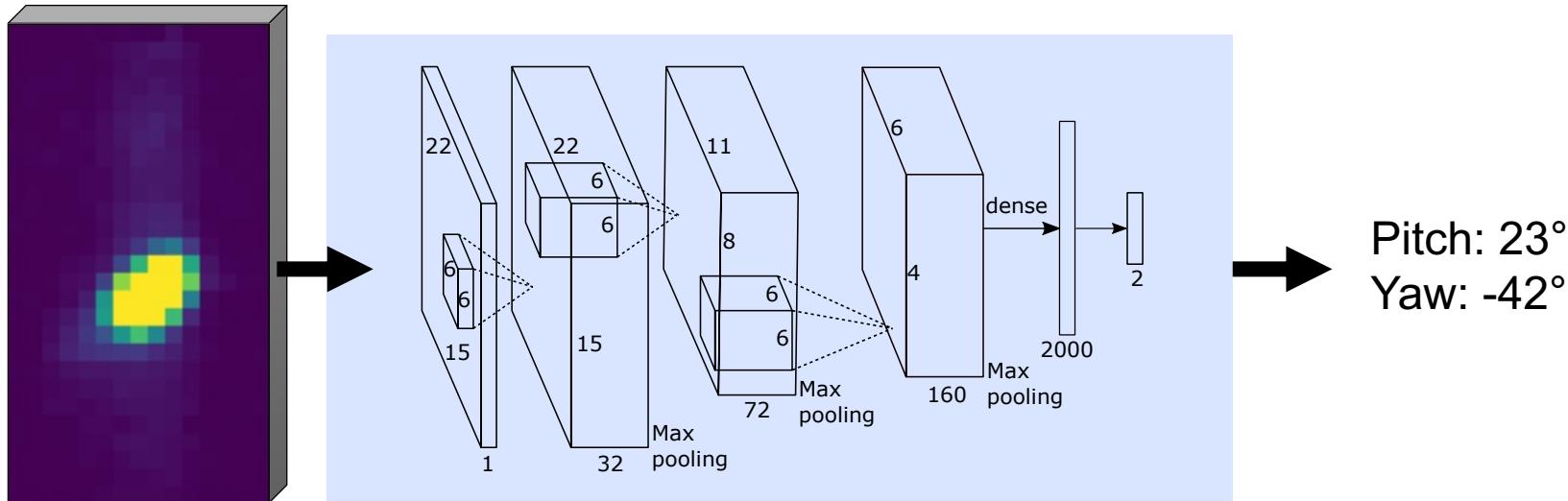
Example

Human-Computer Interaction



Example

Human-Computer Interaction



Method	Pitch			Yaw		
	RMSE	MAE	SD	RMSE	MAE	SD
GP reimplementation of Xiao et al. [41]*	14.74	11.78	14.38	56.58	40.51	39.51
pseudo implementation of Xiao et al. [41]**	14.19	11.58	8.21	44.53	33.39	29.46
CNN + L2	12.8	10.09	7.88	24.19	17.62	16.58

8.9% 45.7%

Neuronal Networks

What can be trained?

Psychological Review
Vol. 65, No. 6, 1958

THE PERCEPTRON: A PROBABILISTIC MODEL FOR INFORMATION STORAGE AND ORGANIZATION IN THE BRAIN¹

F. ROSENBLATT

Cornell Aeronautical Laboratory

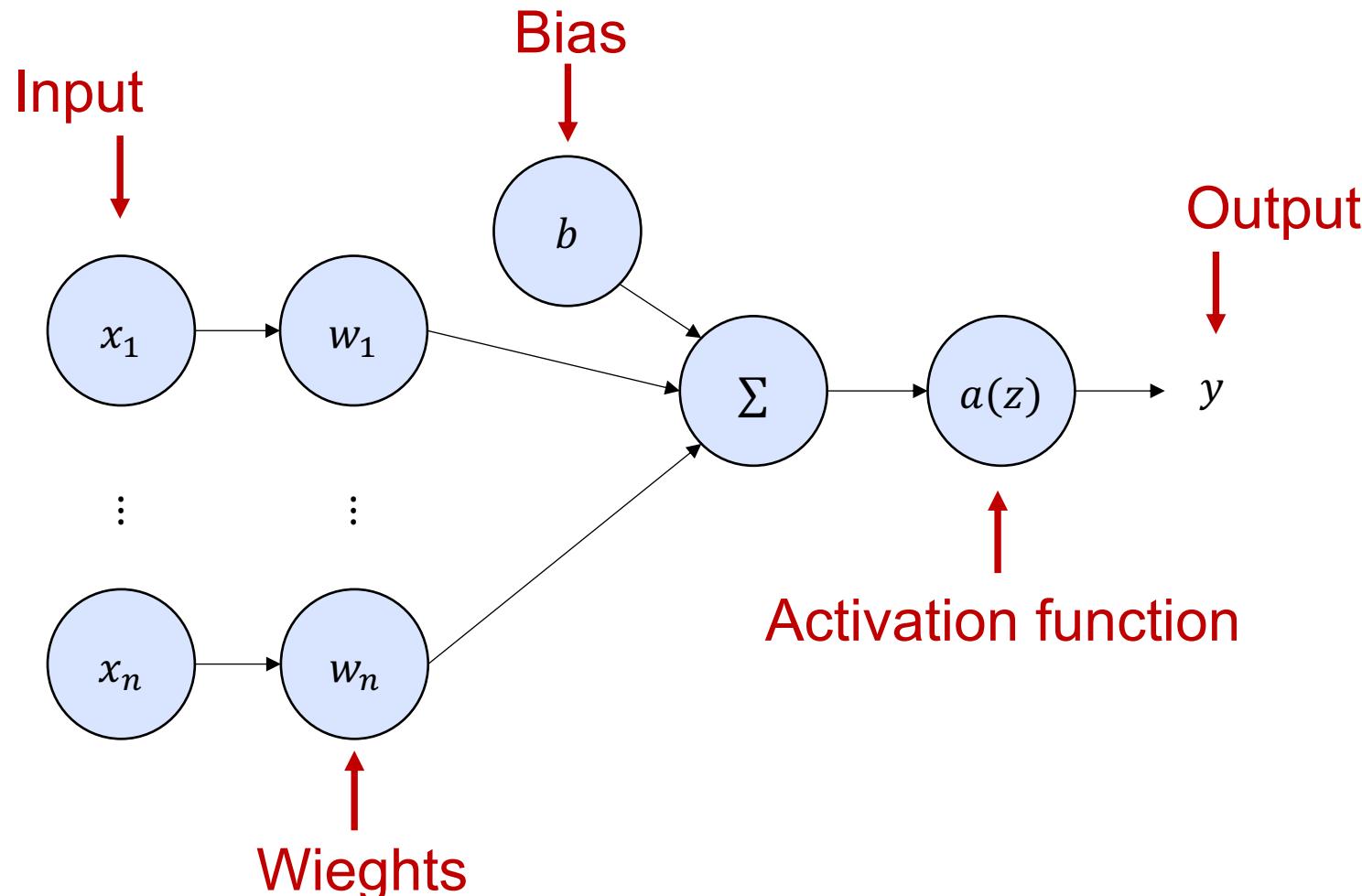
If we are eventually to understand the capability of higher organisms for perceptual recognition, generalization, recall, and thinking, we must first have answers to three fundamental

and the stored pattern. According to this hypothesis, if one understood the code or "wiring diagram" of the nervous system, one should, in principle, be able to discover exactly what an

Frank Rosenblatt. "The perceptron: a probabilistic model for information storage and organization in the brain." *Psychological review* 65, no. 6 (1958): 386. DOI: <https://psycnet.apa.org/doi/10.1037/h0042519>

What is a Perceptron?

Single-Layer Perceptron



Conclusion

Introduction to Machine Learning

- Practical Examples
- General understanding of the model $f(x)$
- Deep Learning Approaches
 - Perceptron
 - Weights & biases can be trained
 - Activation function

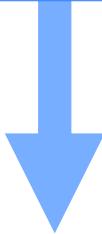
Questions?

Tutorials

Tutorials

- 06.05.2022 Organization & Exercise 01: Recording your own data (2 weeks)
- 13.05.2022 Live Coding Session: Getting Started with Neuronal Networks
- 20.05.2022 Live Coding Session: Deploying Models to Mobile Devices (Android) & Exercise 02: Clearing your data and training the first model (2 weeks)
- 27.05.2022 canceled
- 03.06.2022 Project Ideation & Exercise 03: Training an improved model based on a large dataset (1 weeks)
- 10.06.2022 Project Pitches: Show Current Project Status
- 17.06.2022 Individual Help for Projects
- 24.06.2022 Individual Help for Projects
- 01.07.2022 Individual Help for Projects
- 08.07.2022 Individual Help for Projects
- 15.07.2022 canceled
- 22.07.2022 How to give a great project presentation & Q'n'A: Exam preparation & Individual Help for Projects
- 29.07.2022 Final Presentation

Projects

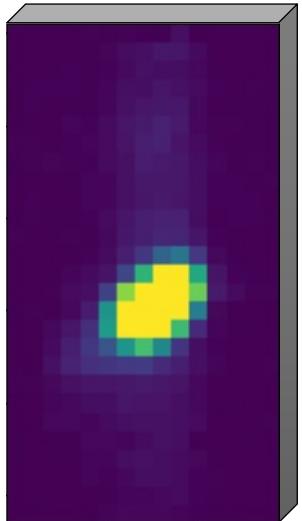


Tutorials

- Exercise 01: Recording your own data
 - 2 weeks
- Exercise 02: Clearing your data and training the first model
 - 2 weeks
- Exercise 03: Training an improved model based on a large dataset
 - 1 week

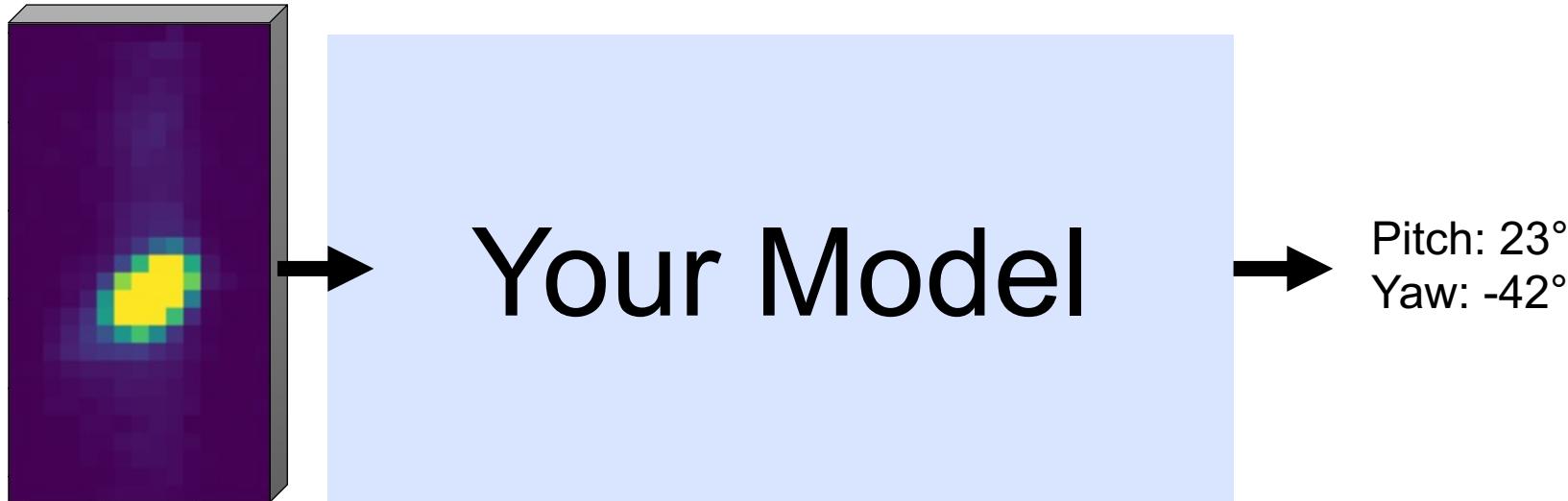
Example

Human-Computer Interaction



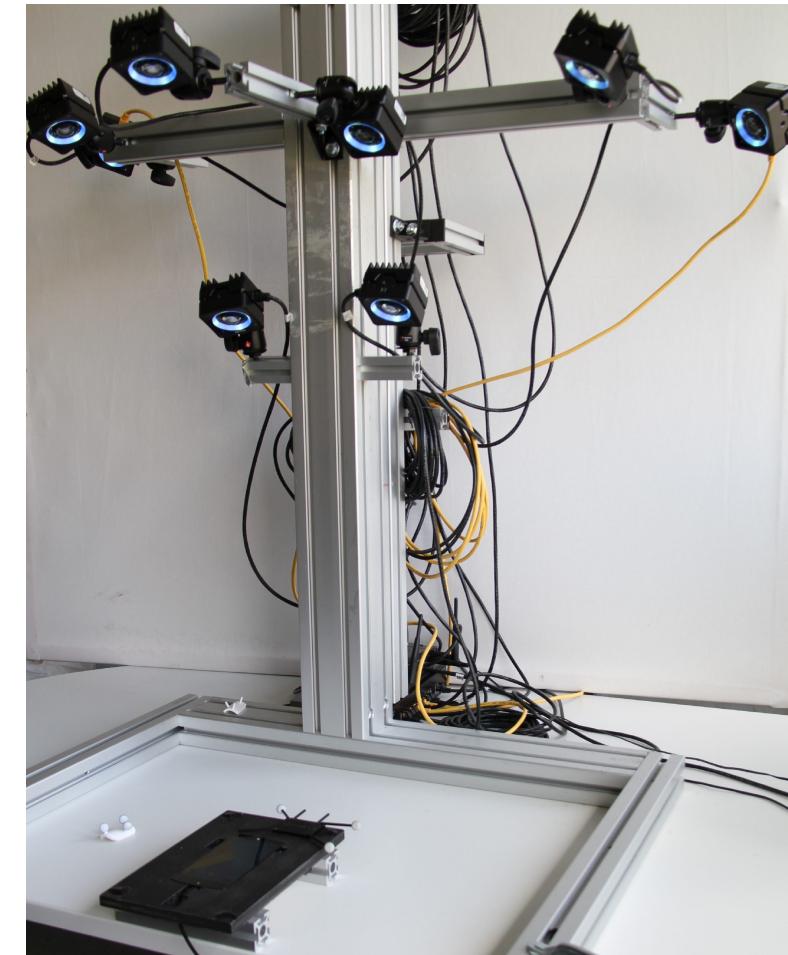
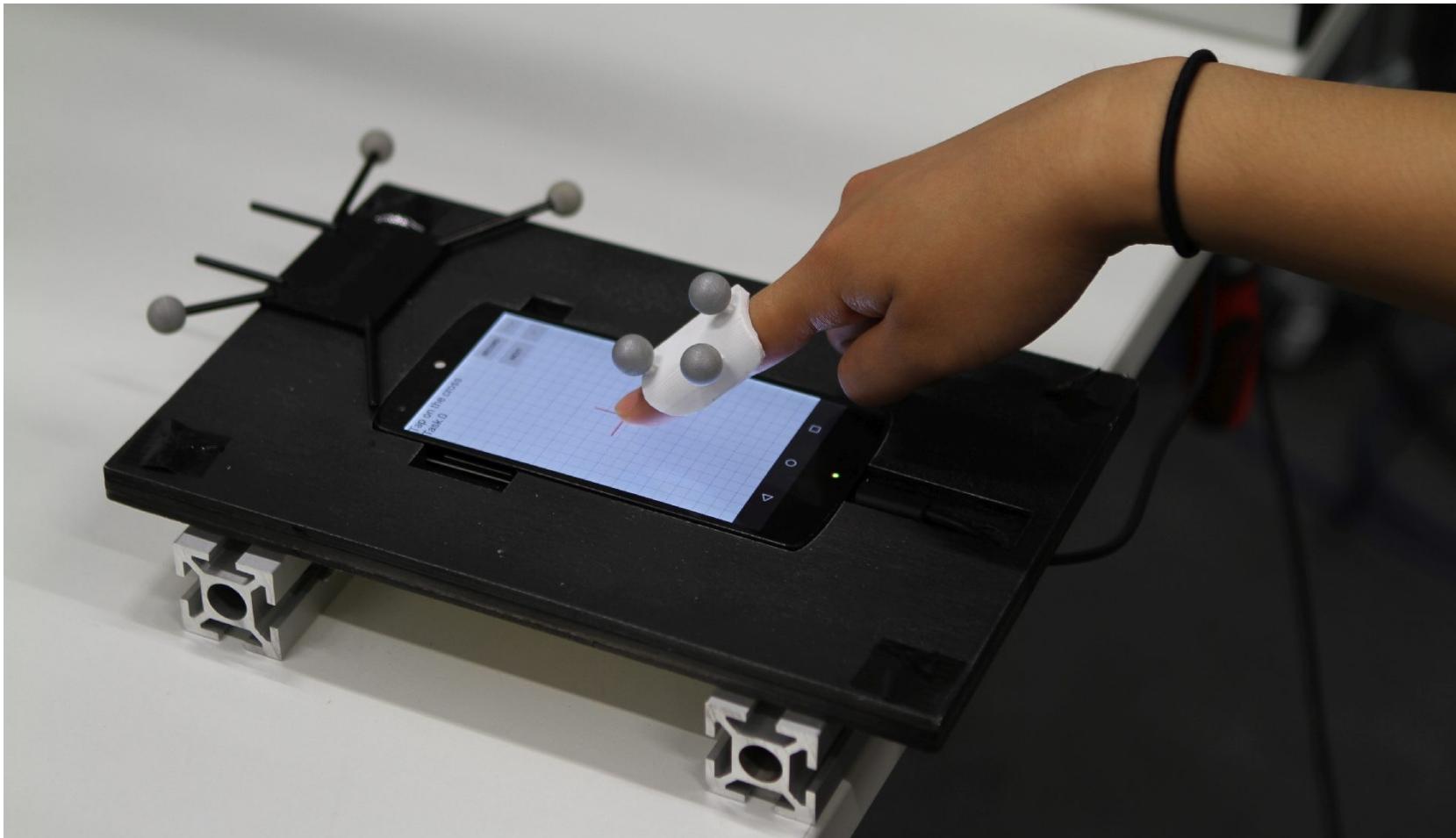
Example

Human-Computer Interaction



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Exercise 01: Recording Your Own Data



Exercise 01: Recording Your Own Data

- You are required to record your own dataset
- Location:
 - Frauenlobstraße 7A
80337 München

Detailed Information will flow by Monday



Exercise 02: Data and Training

- **Step 1: Clean your data**
 - Start here: https://github.com/interactionlab/Capacitive-Finger-Orientation-Estimation/blob/master/Step1_Preprocessing.ipynb
 - A) This is not enough for data cleaning
 - Hints: clock sync, data augmentation, normalization
 - B) You can be better
- **Step 2: Train your first model**
 - Start here: https://github.com/interactionlab/Capacitive-Finger-Orientation-Estimation/blob/master/Step4_Training_using_Keras.ipynb
 - Goal: outperform Pitch RMSE = 24.19° and Yaw RMSE = 12.80°
 - Note: due to the not diverse dataset this should be easily possible.

Exercise 03: Improve your Model

- We will collect all of your data, preprocess it, and merge it.
- => 50 datasets
- => 50 times the diversity
- => 50 times more robust – but way harder to train
- **Goal:** Outperform Pitch RMSE = 24.19° and Yaw RMSE = 12.80°

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Sven Mayer