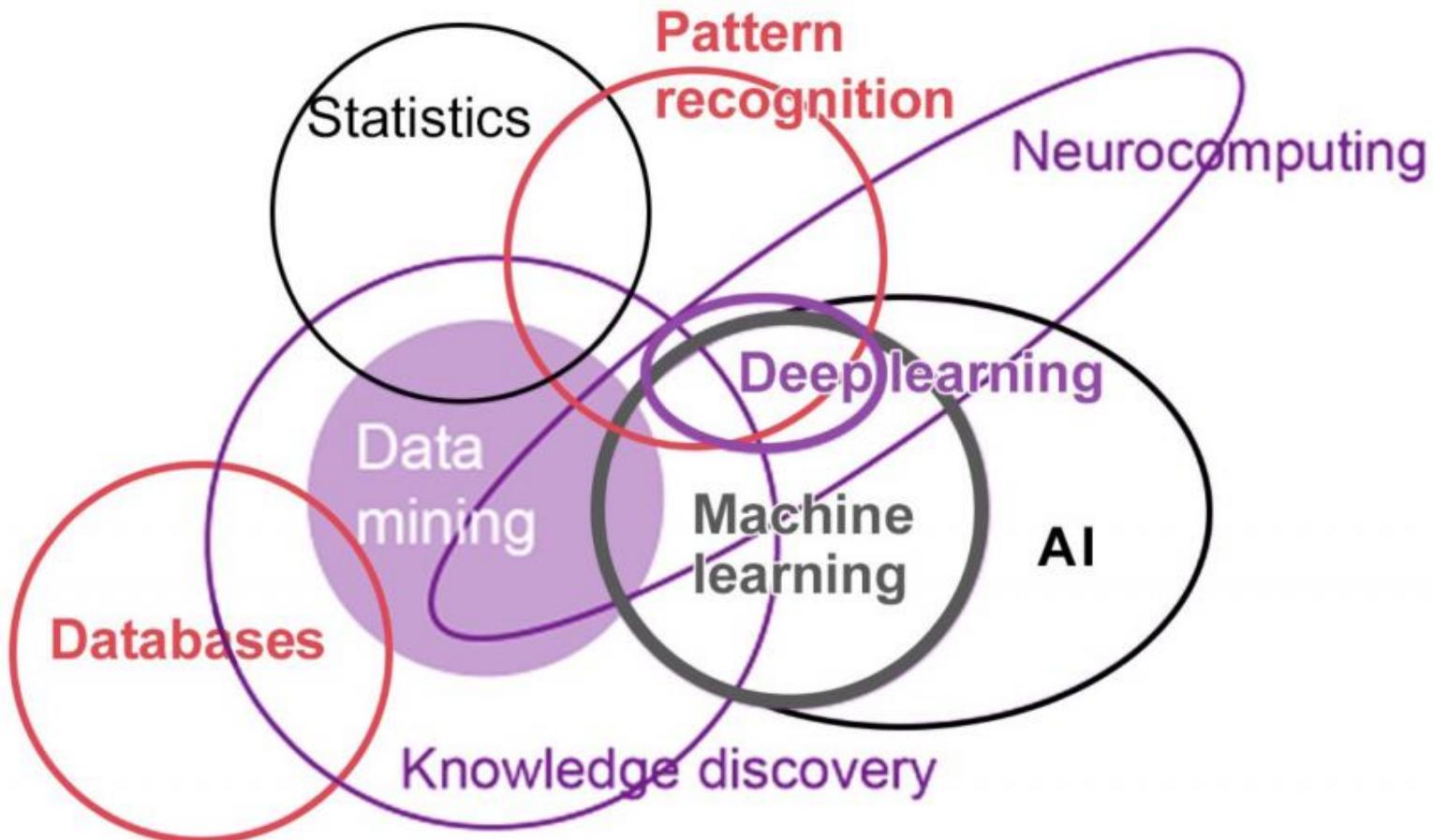


# Machine learning at the edge

---

I NYOMAN KUSUMA WARDANA

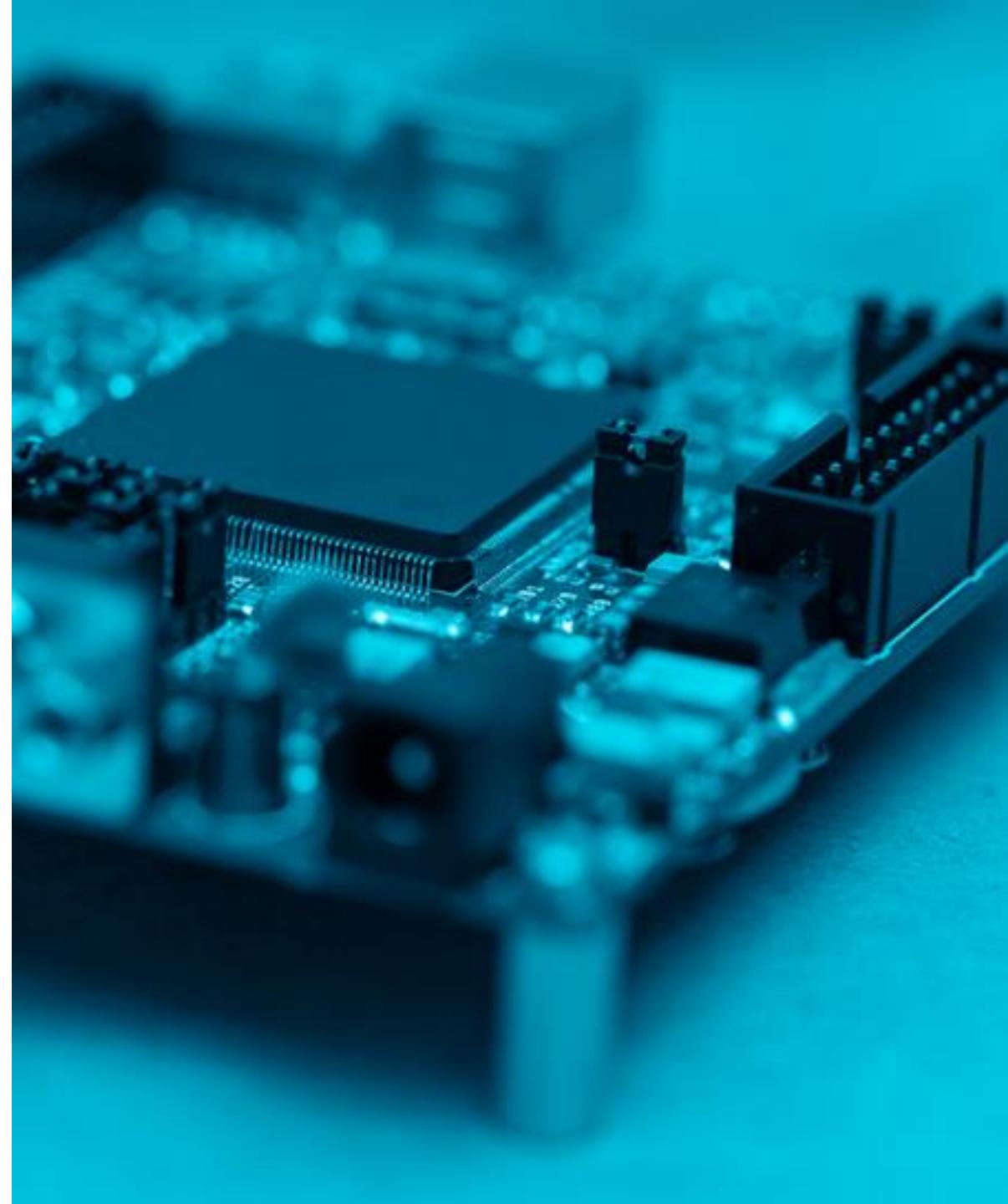


Referensi:

<https://jaylatta.net/history-of-ai-from-winter-to-winter/>

---

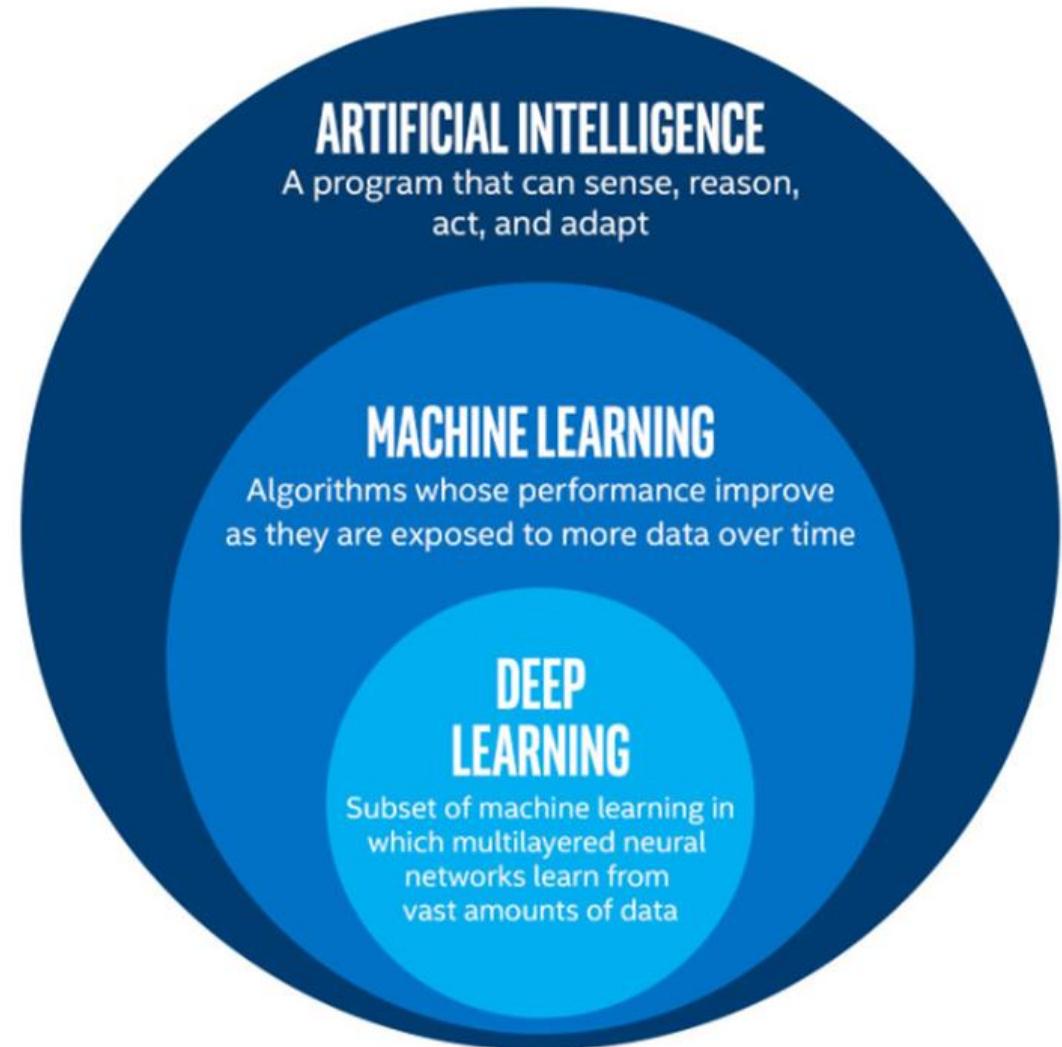
Perkenalkan saya dengan  
beberapa terminologi  
terkait ML ...



**Machine learning** adalah bagian dr artificial intelligence (AI) yang secara umum didefinisikan sebagai **kemampuan mesin untuk meniru perilaku manusia**.

**Machine learning** adalah salah satu cara untuk mewujudkan AI. Istilah ML dipopulerkan oleh **Arthur Samuel** pada akhir tahun 1950-an, yg mendefinisikan ML sebagai “**bidang studi yang memberi komputer kemampuan untuk belajar tanpa diprogram secara eksplisit**”

Machine learning dari **data** — angka, gambar, teks, dsb.

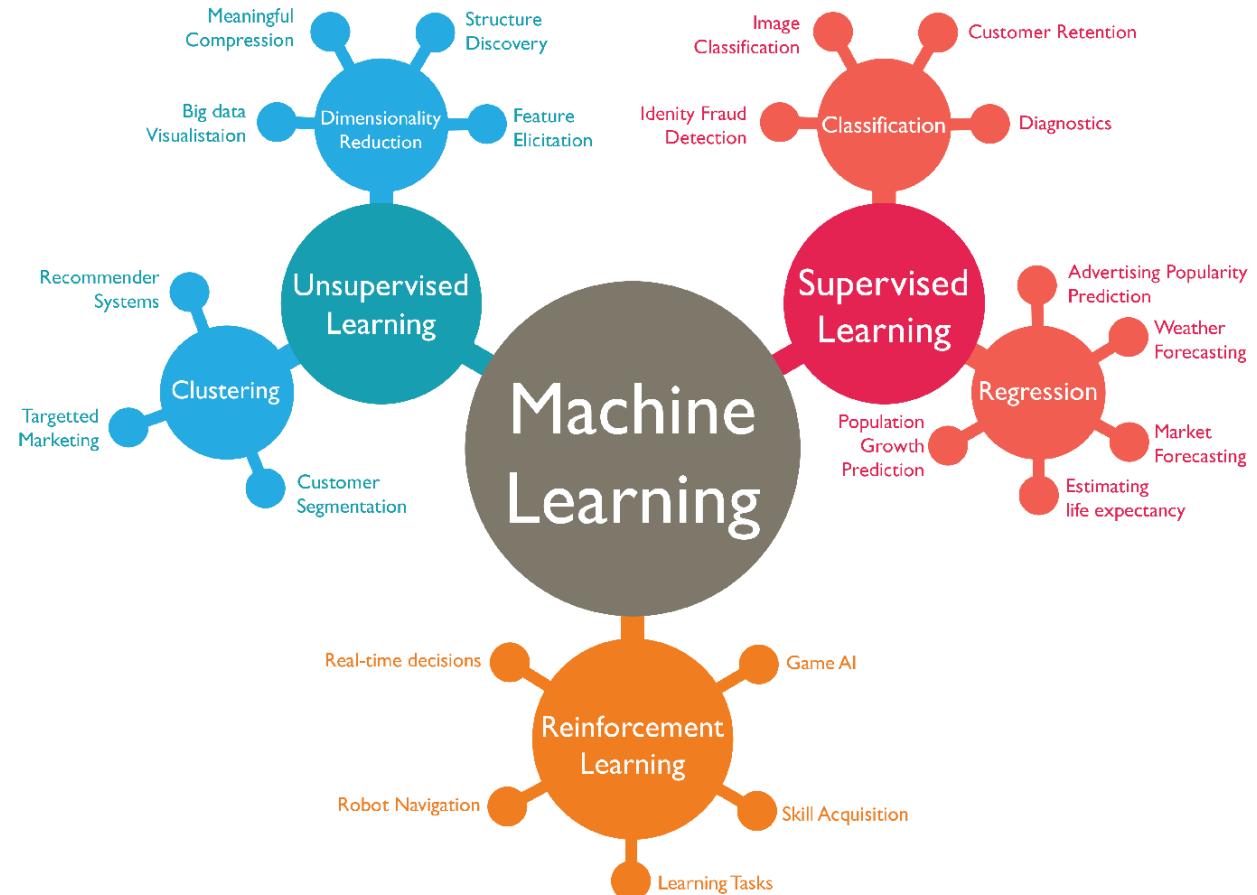


*Referensi:*

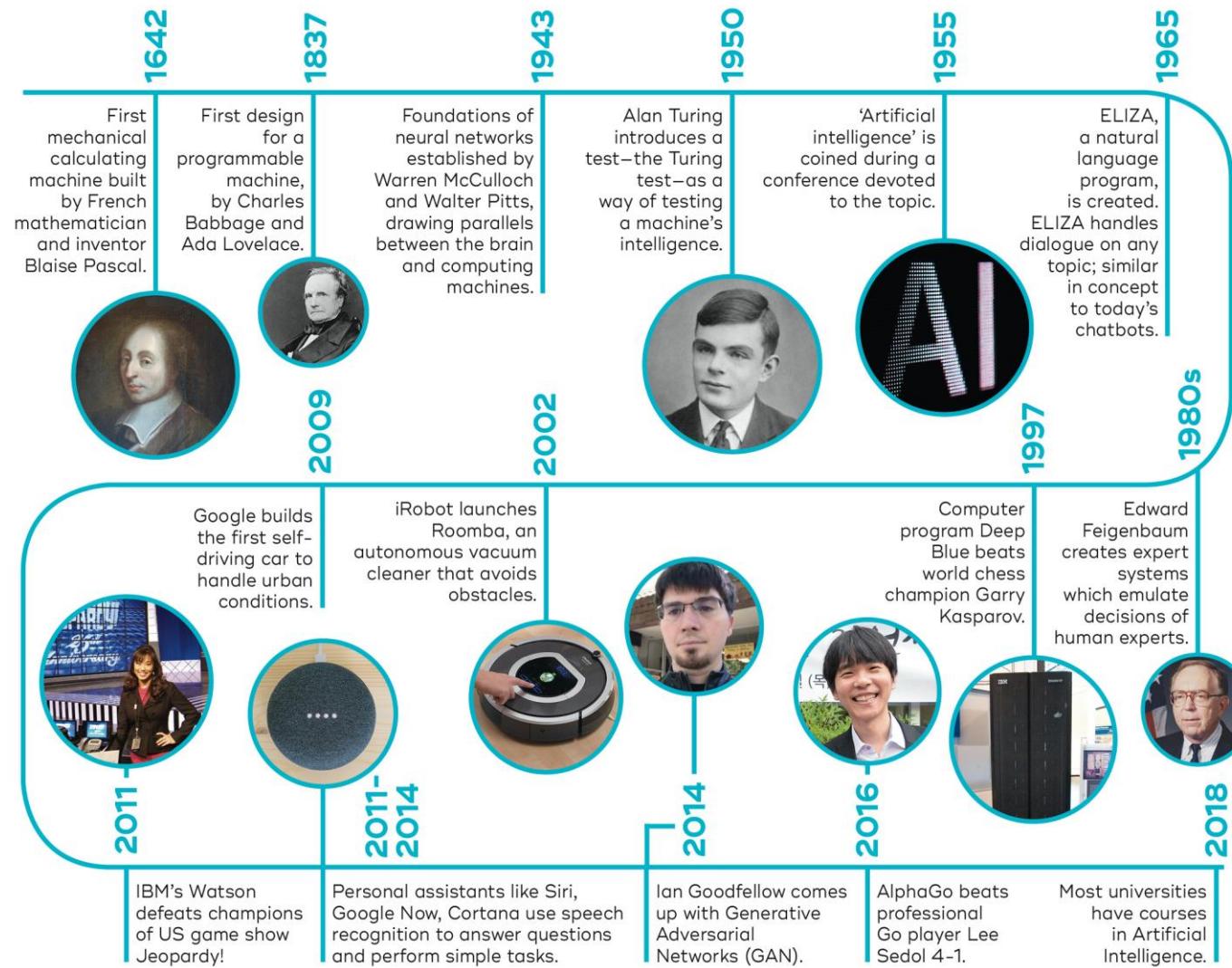
<https://mitsloan.mit.edu/ideas-made-to-matter/machine-learning-explained>

## Penerapan Sistem Berbasis Machine Learning:

- **Descriptive** – sistem menggunakan data untuk menjelaskan apa yang telah terjadi
- **Predictive** – sistem menggunakan data untuk memprediksi apa yang akan terjadi
- **Prescriptive** – sistem akan menggunakan data untuk membuat saran tentang apa yang akan dilakukan



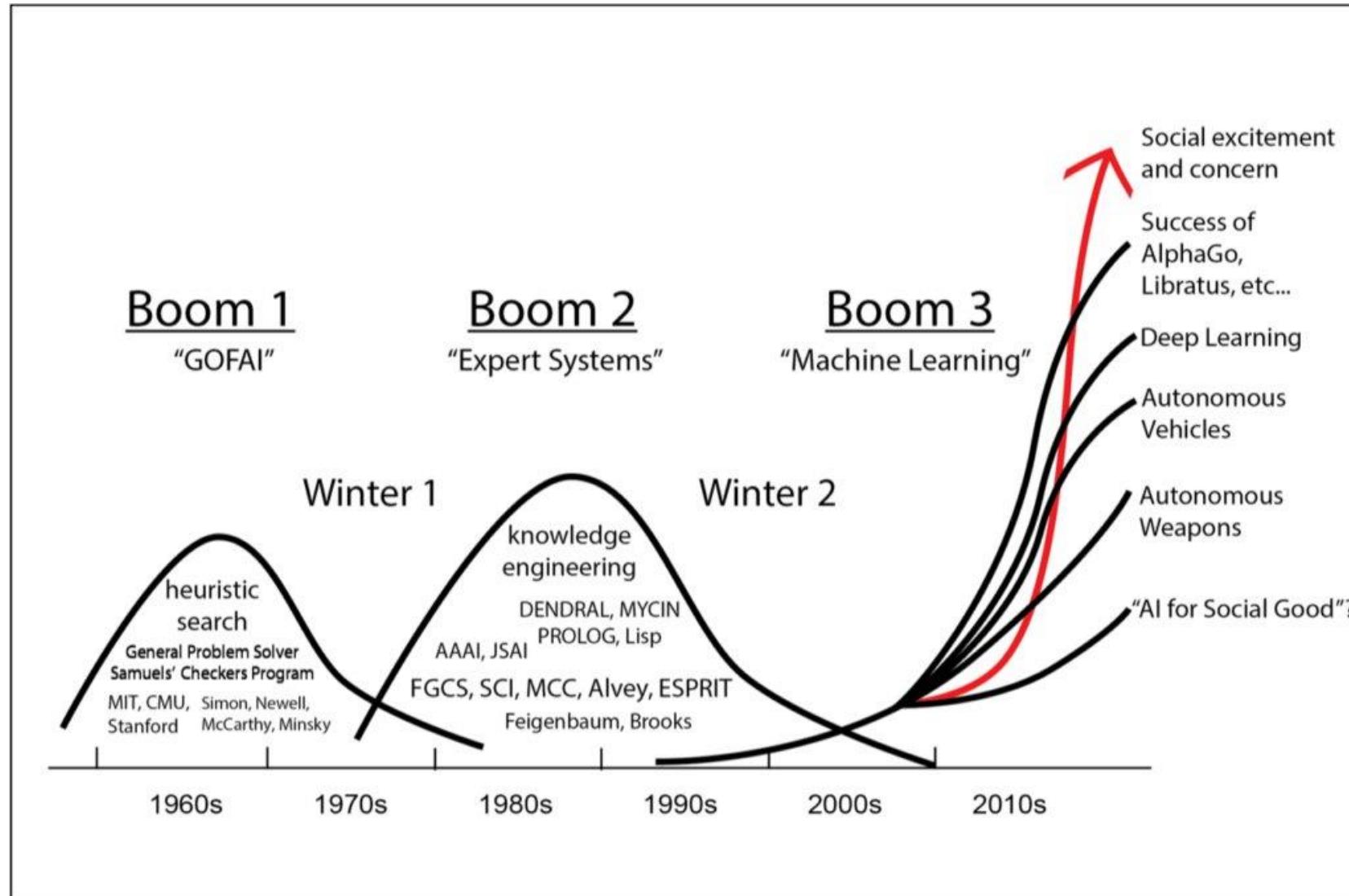
# Sejarah AI



Referensi:

<https://qbi.uq.edu.au/brain/intelligent-machines/history-artificial-intelligence>

# AI: Winter ke Winter

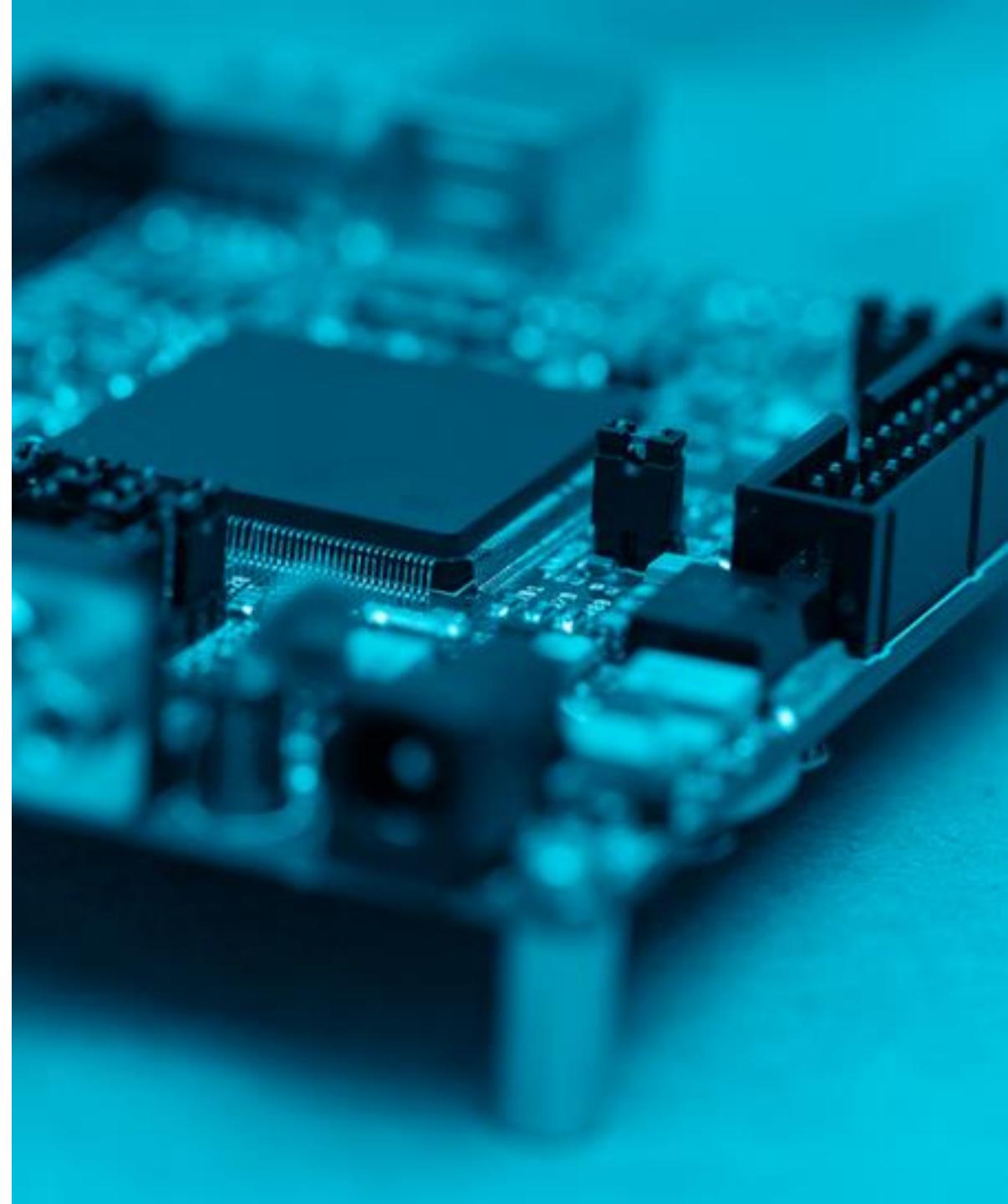


Referensi:

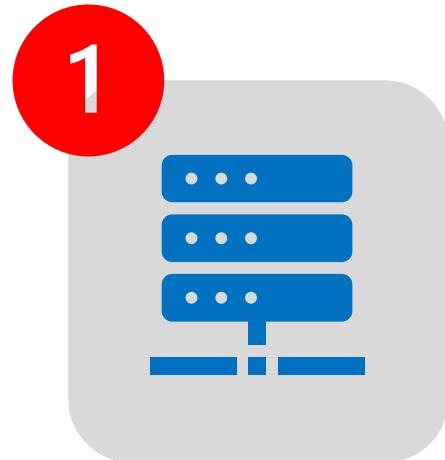
<https://jaylatta.net/history-of-ai-from-winter-to-winter/>

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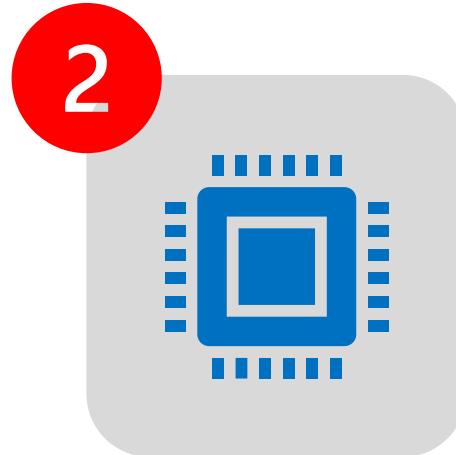
Apa yang membuat ML  
menjadi begitu popular  
saat ini?



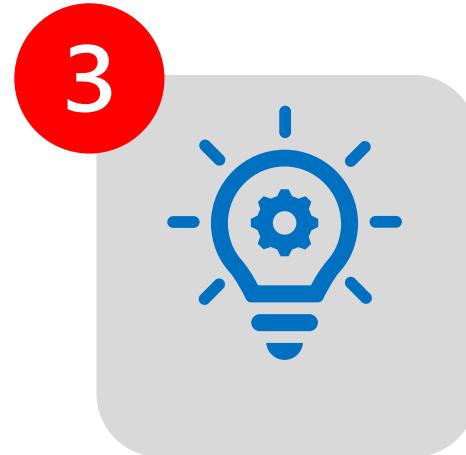
# Faktor pendorong berkembangnya ML



Ketersediaan  
Data



Kemampuan  
Komputasi



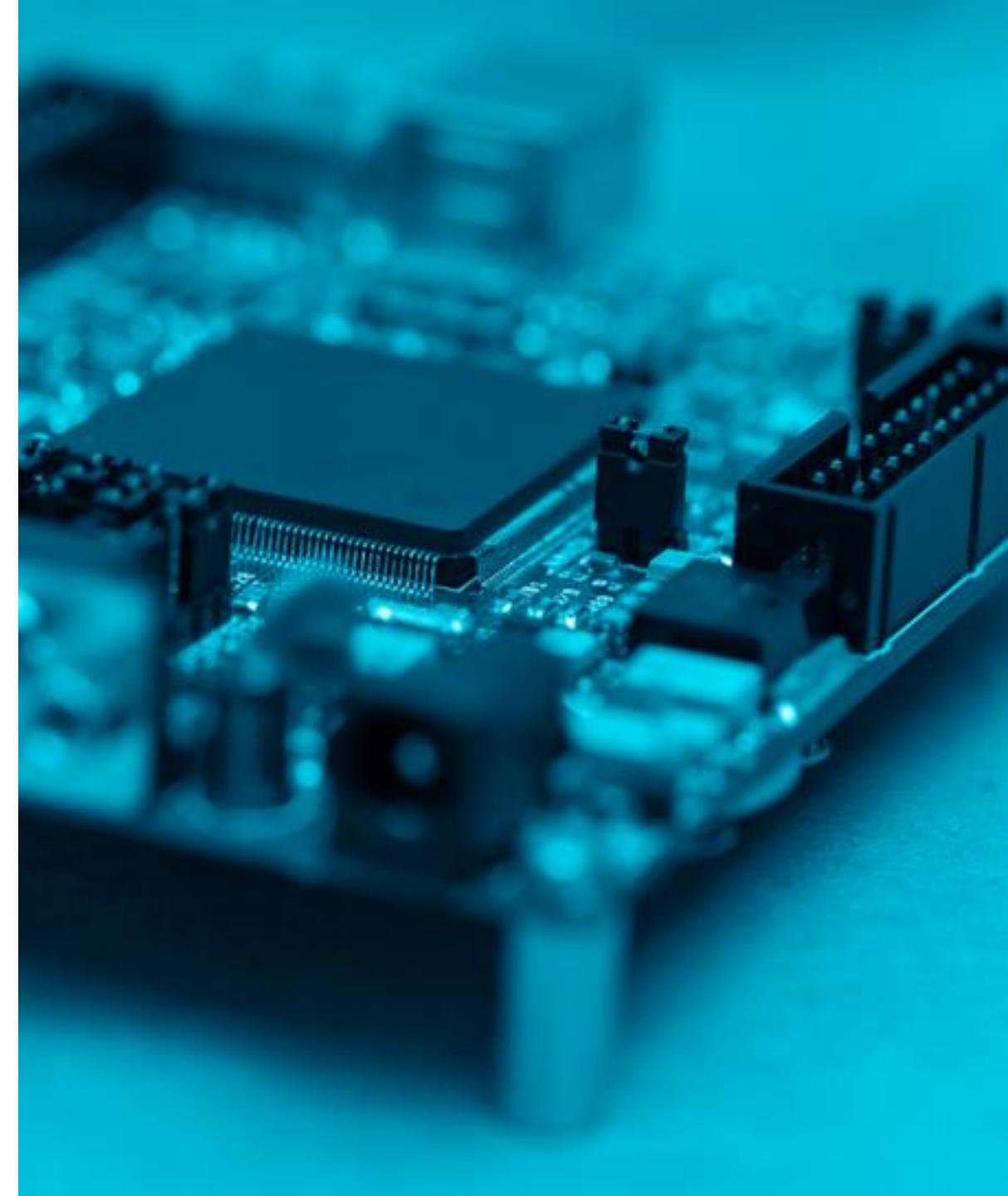
Perkembangan  
Algoritma



Kepentingan  
Publik

---

Trus... saya bisa mulai  
belajar ML dari mana?



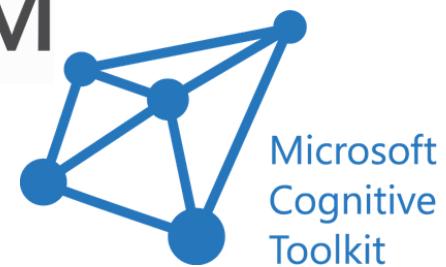
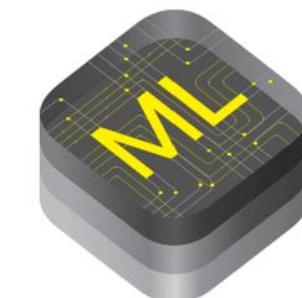
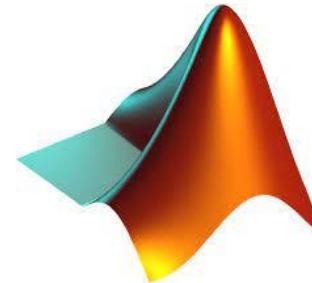
# Popular Machine Learning Frameworks



TensorFlow

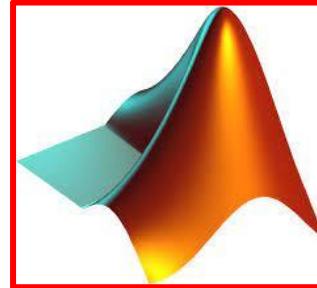


將軍

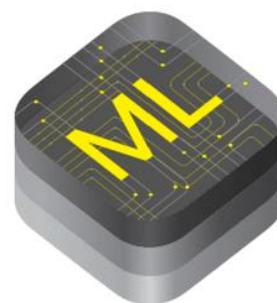


---

# Anda bisa memulai dari ini:

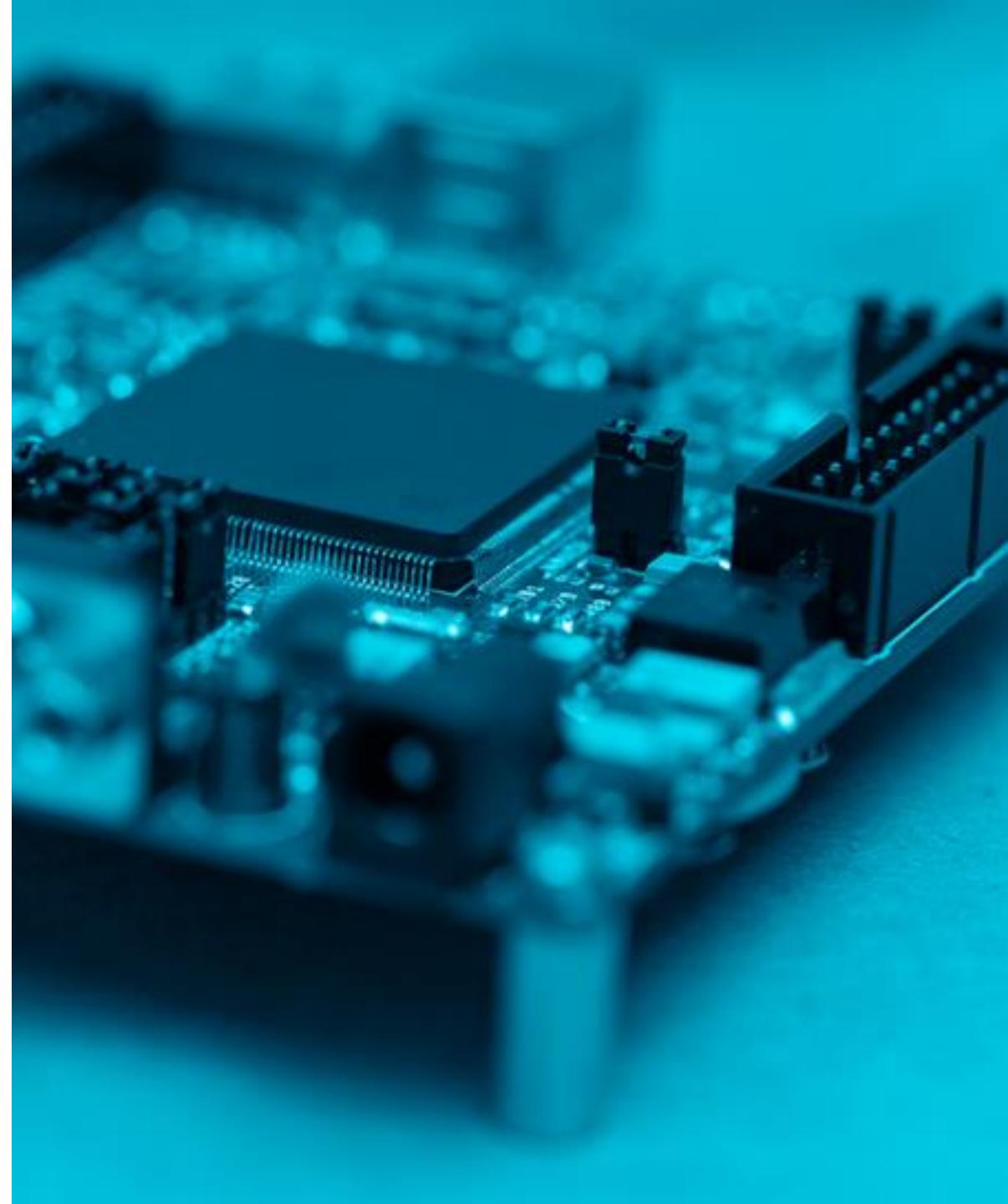


將軍

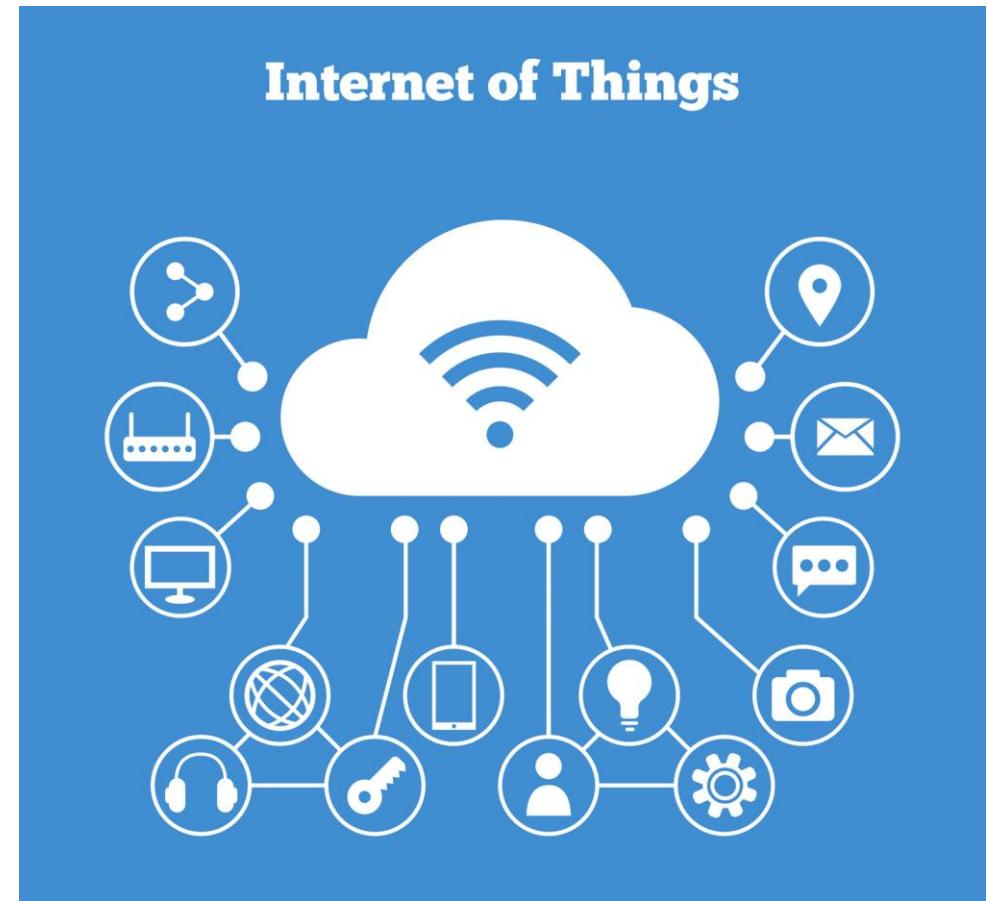
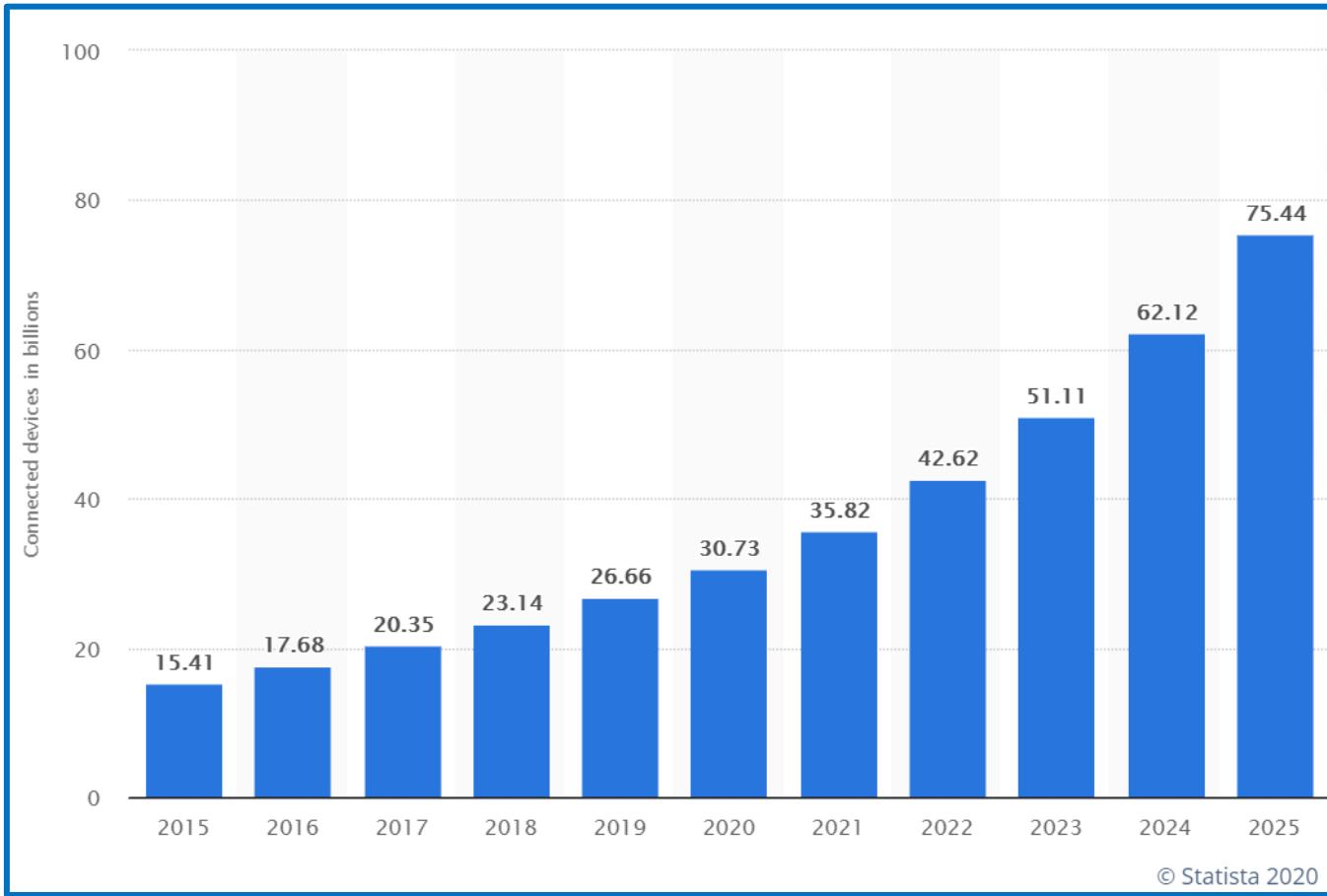


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# Apa itu Cloud Computing dan Edge Computing?

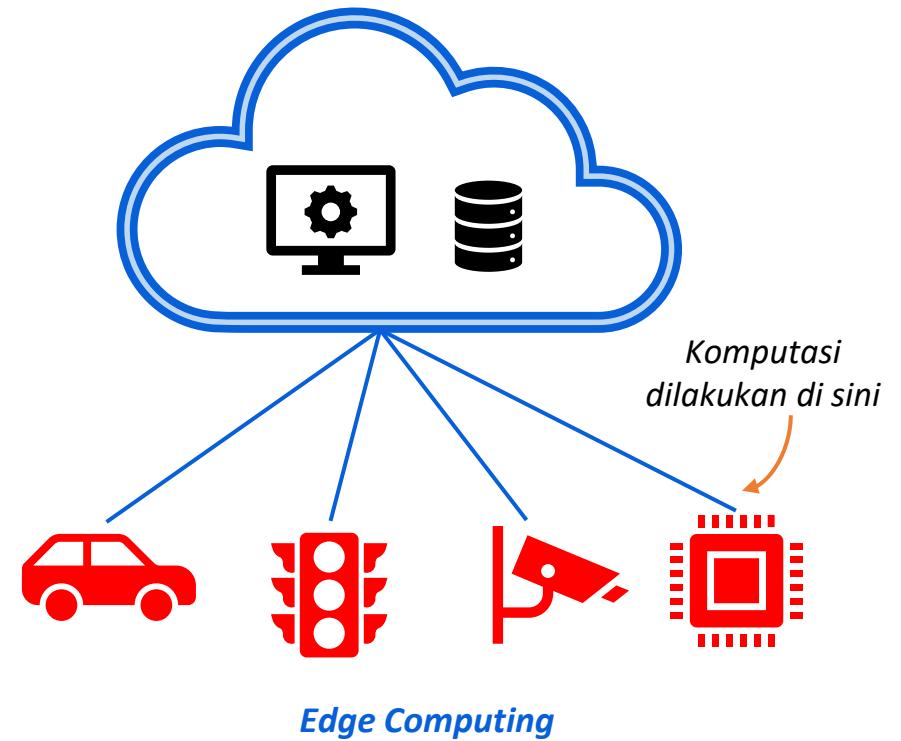
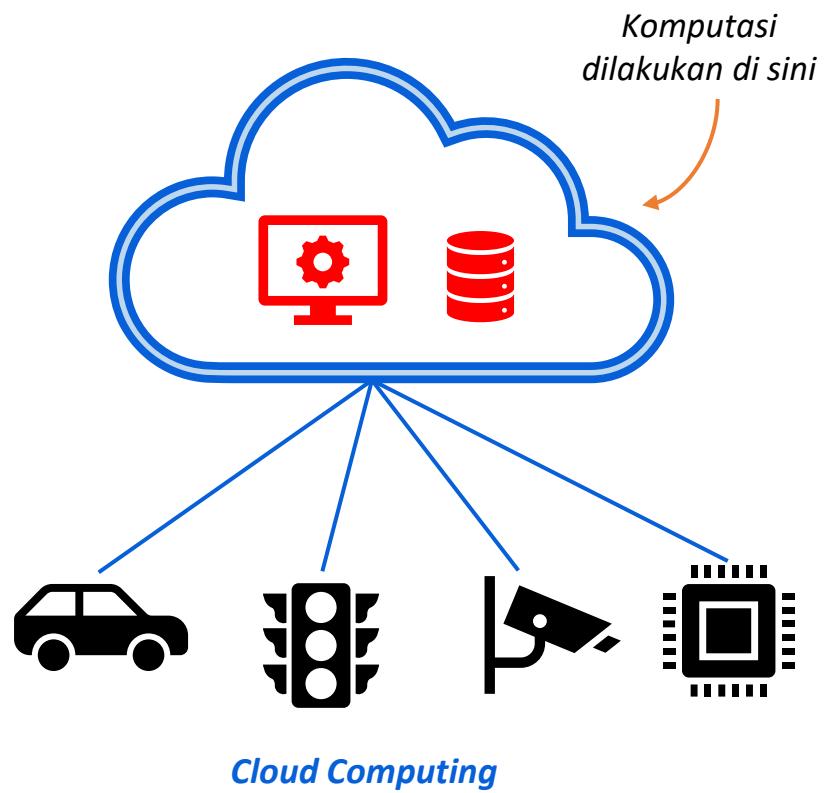


## Prediksi Jumlah Alat yang Terkoneksi Internet of Things (IoT) dari thn. 2015 - 2025



The total installed base of Internet of Things (IoT) connected devices is projected to amount to 75.44 billion worldwide by 2025, a fivefold increase in ten years. The IoT, enabled by the already ubiquitous Internet technology, is the next major step in delivering Internet's promise of making the world a connected place. (statistica.com)

# Edge computing vs. cloud computing



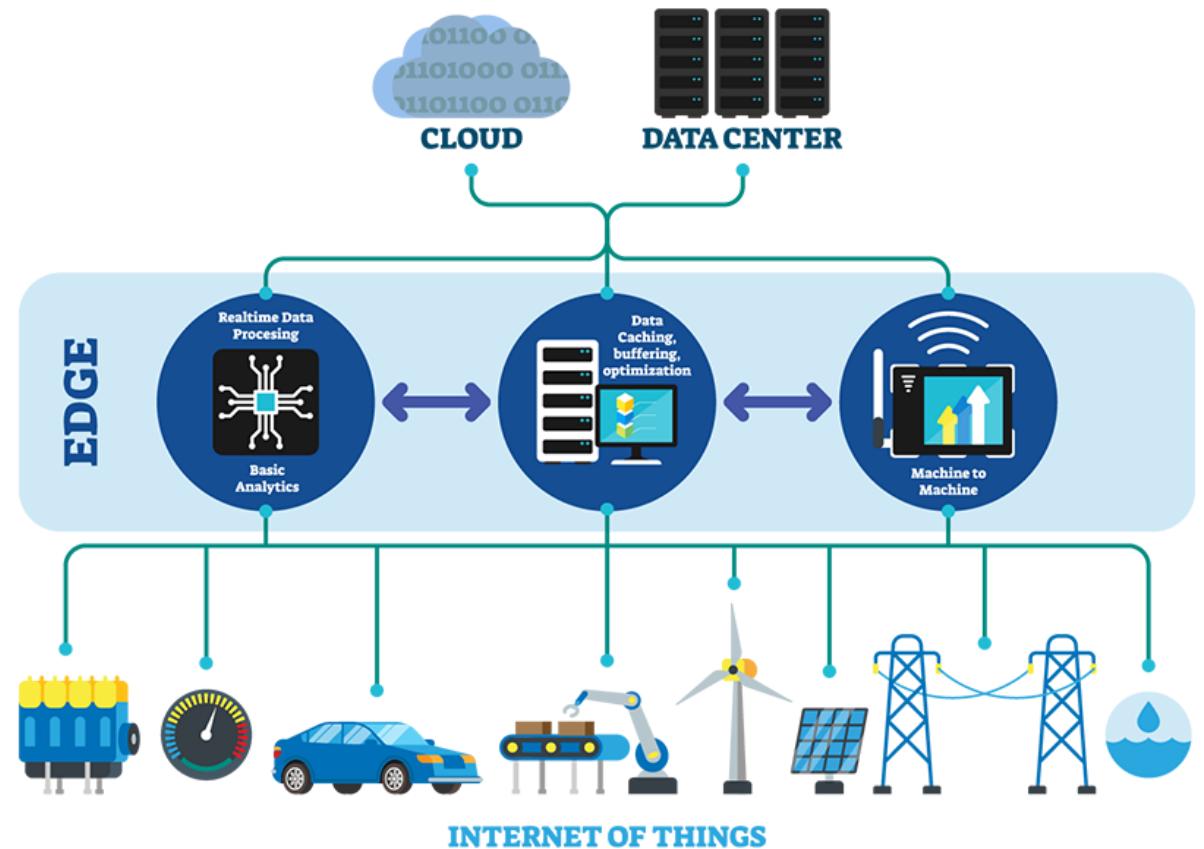
## • Tantangan Cloud Computing:

- Isu keamanan data
- Masalah performa
- Biaya operasional

## Keuntungan Edge Computing:

- Meningkatkan privasi dan keamanan data
- Performa lebih responsif dan tangguh
- Mengurangi biaya operasional
- Meningkatkan reliability dan efisiensi bisnis
- Unlimited scalability
- Mengurangi latensi

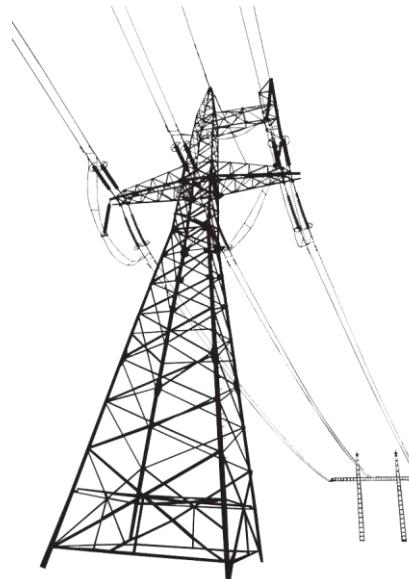
# Edge Computing



Referensi:

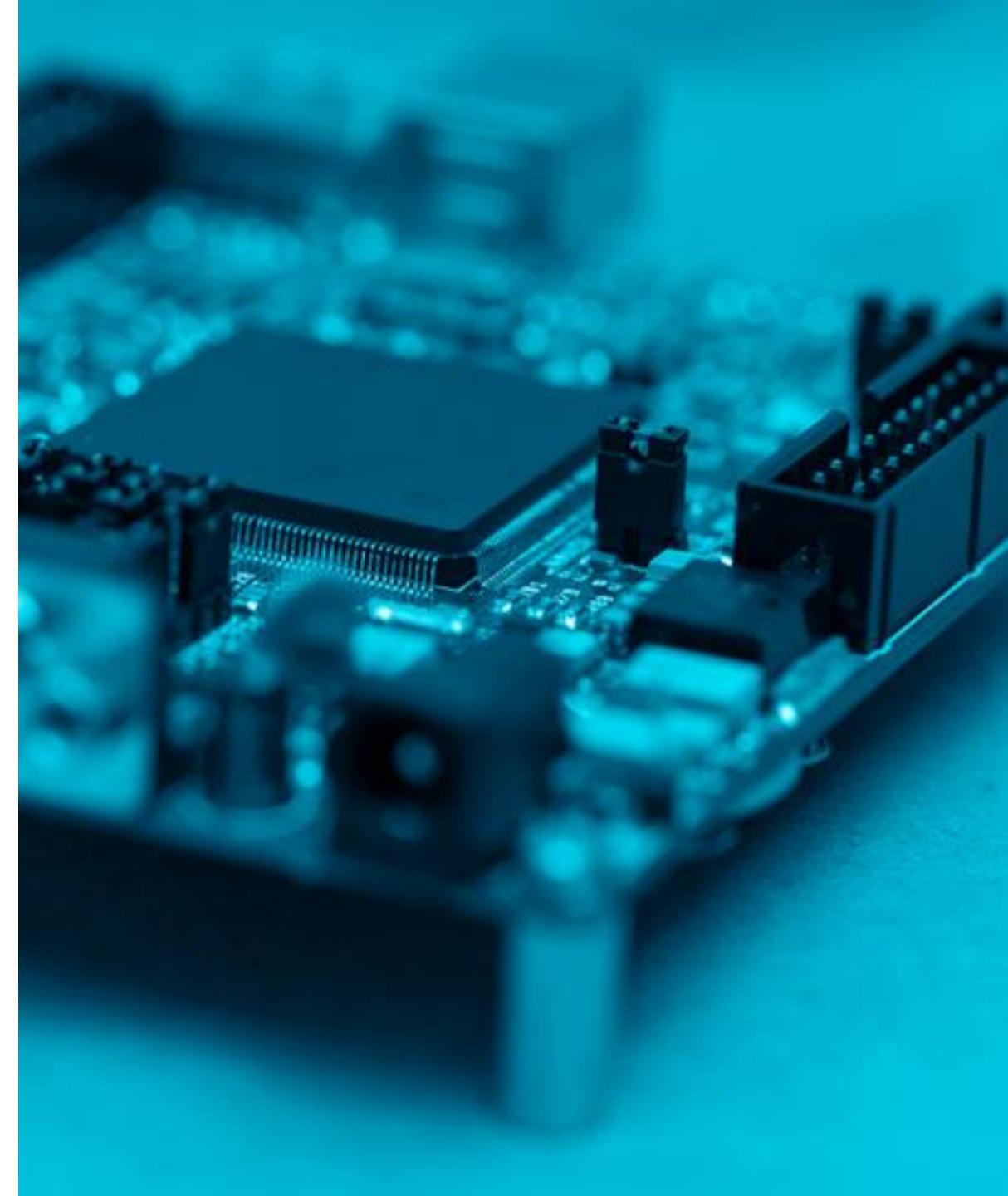
<https://innovationatwork.ieee.org/real-life-edge-computing-use-cases/>

# Edge Devices



---

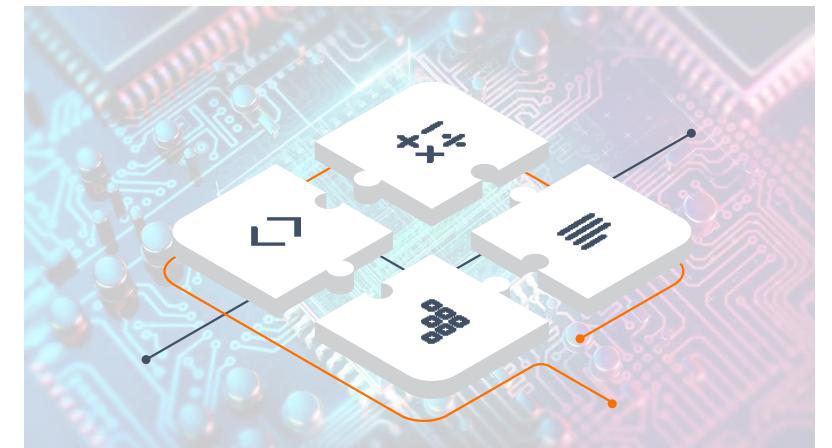
Apa itu embedded systems?



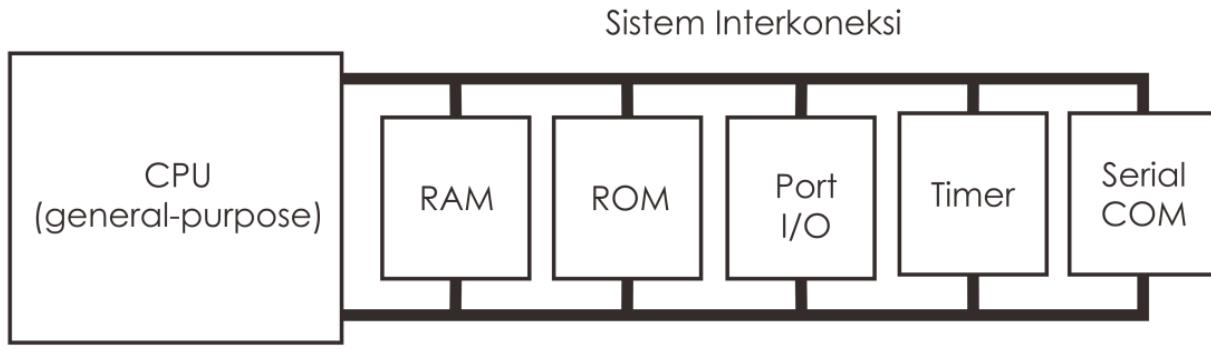
# Embedded Systems

- **System** → A system is an arrangement in which all its units **assemble work together** according to a set of rules. It can also be defined as **a way of working, organizing or doing one or many tasks** according to a fixed plan.
- **Embedded System** → Embedded means something that is **attached to another thing**. An embedded system can be thought of as a computer hardware system having software embedded in it. An embedded system is a microcontroller or microprocessor-based system which is designed to **perform a specific task**.
- **Characteristics** of an Embedded System:

- Single-functioned
- Tightly constrained
- Reactive and Real time
- Microprocessors based
- Memory
- Connected
- HW-SW systems



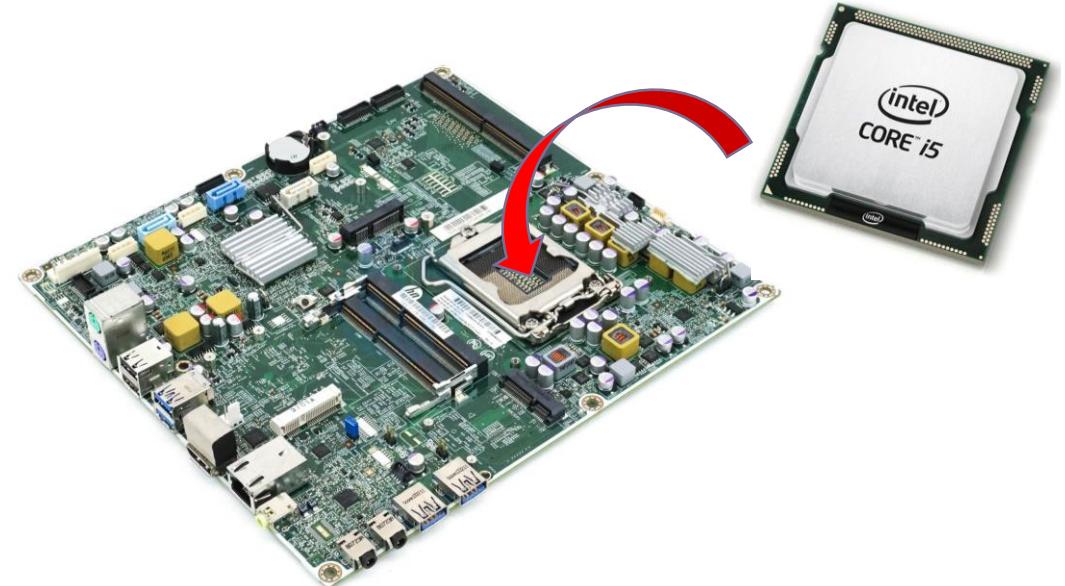
# General purpose processor vs Microcontroller



(a) Sistem Mikroprosesor

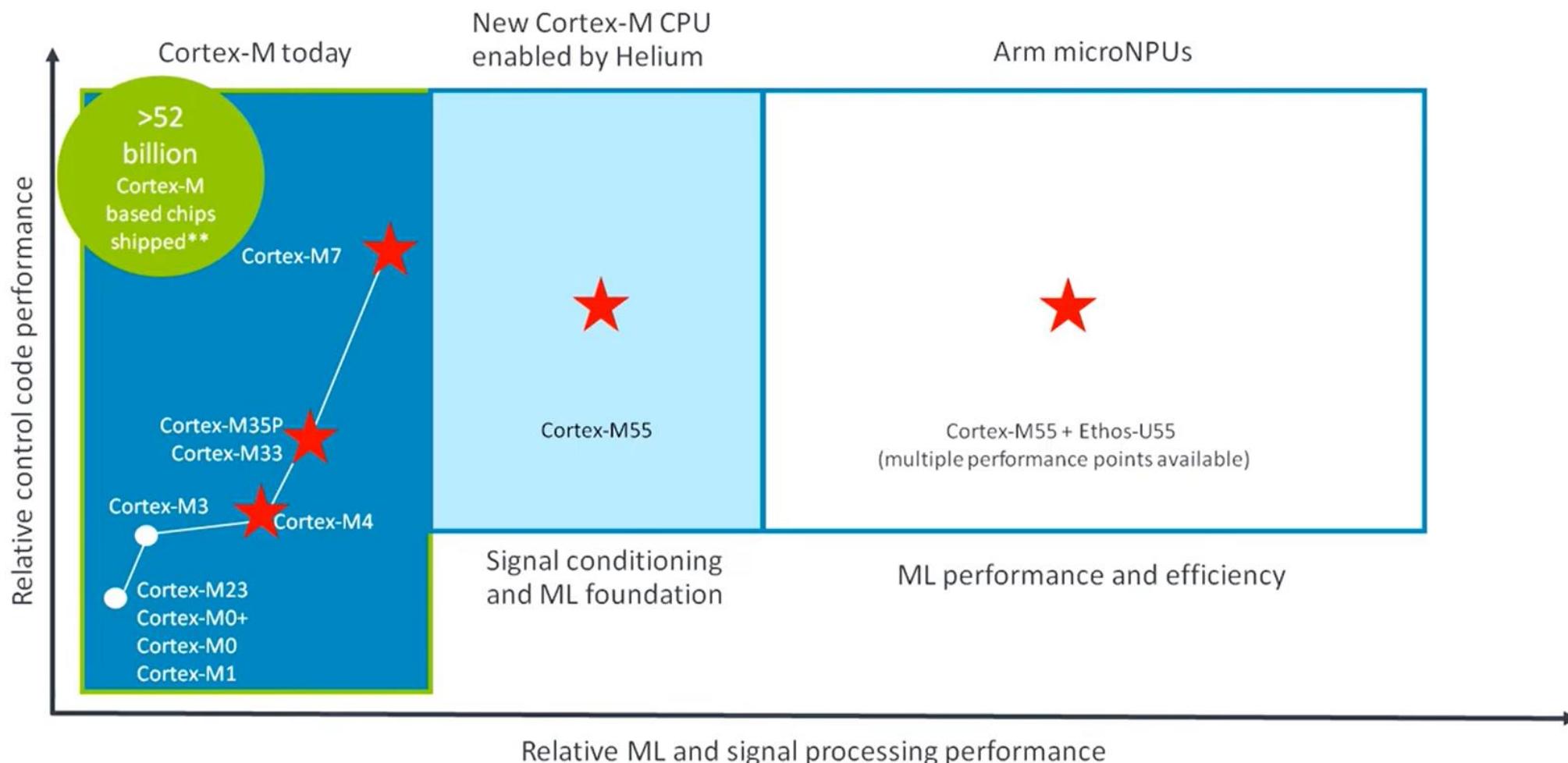
CPU	RAM	ROM
I/O ADC	Timer	Serial COM

(b) Sistem Mikrokontroler



Banyak digunakan di  
penerapan embedded

# Pushing the Boundaries for Real-time On-device Processing



\*\*Based on Arm data

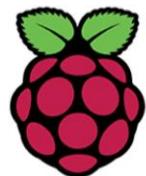
arm

# The future of ML shifts to the edge

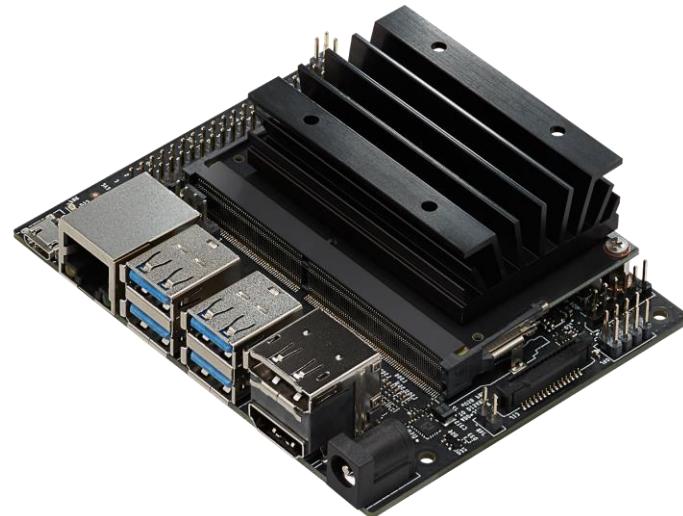
New survey reveals how developers  
are accelerating tinyML innovation



# Bagaimana dengan SBC?



Raspberry Pi



nVIDIA®

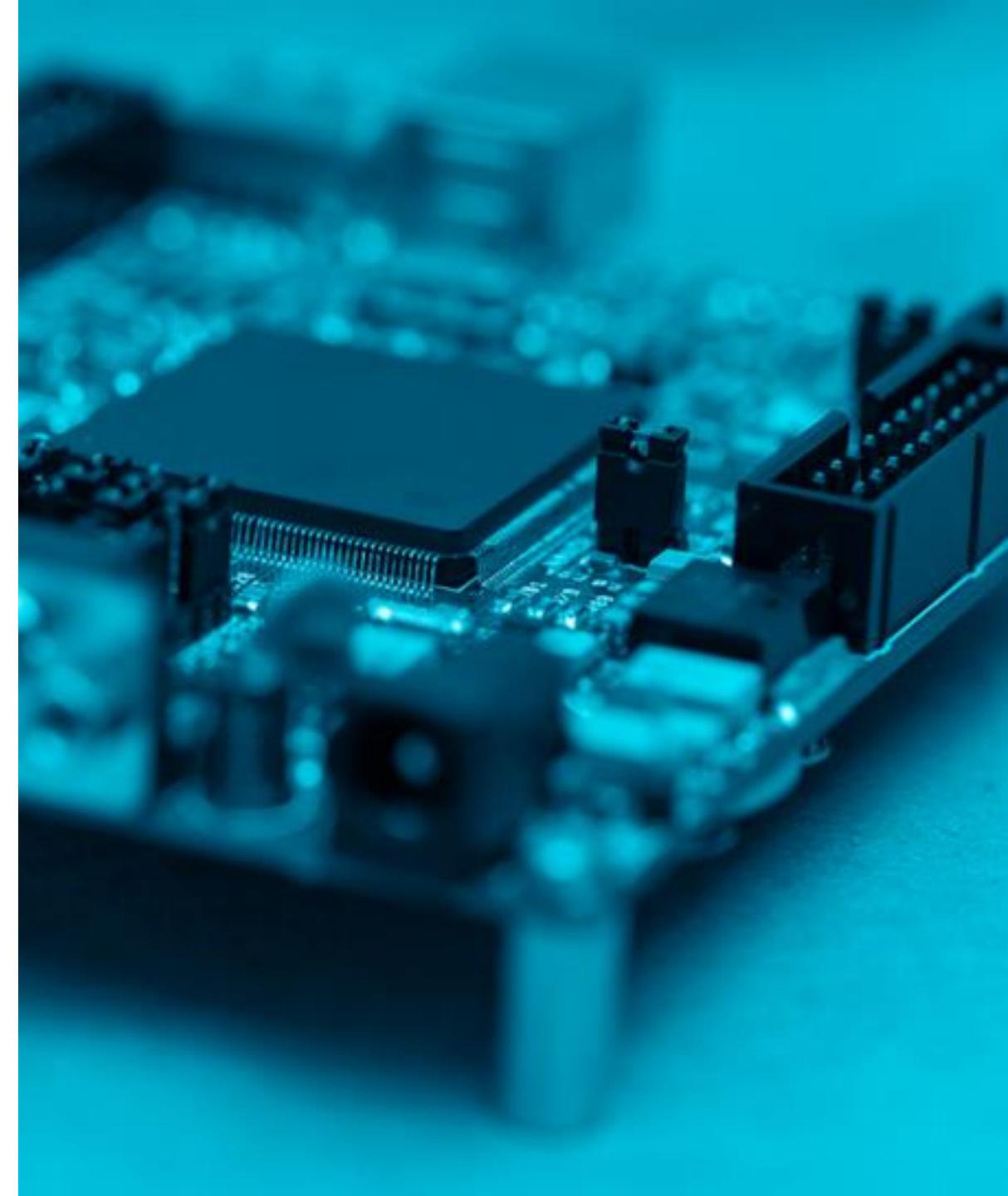
# Bagaimana dengan SBC?



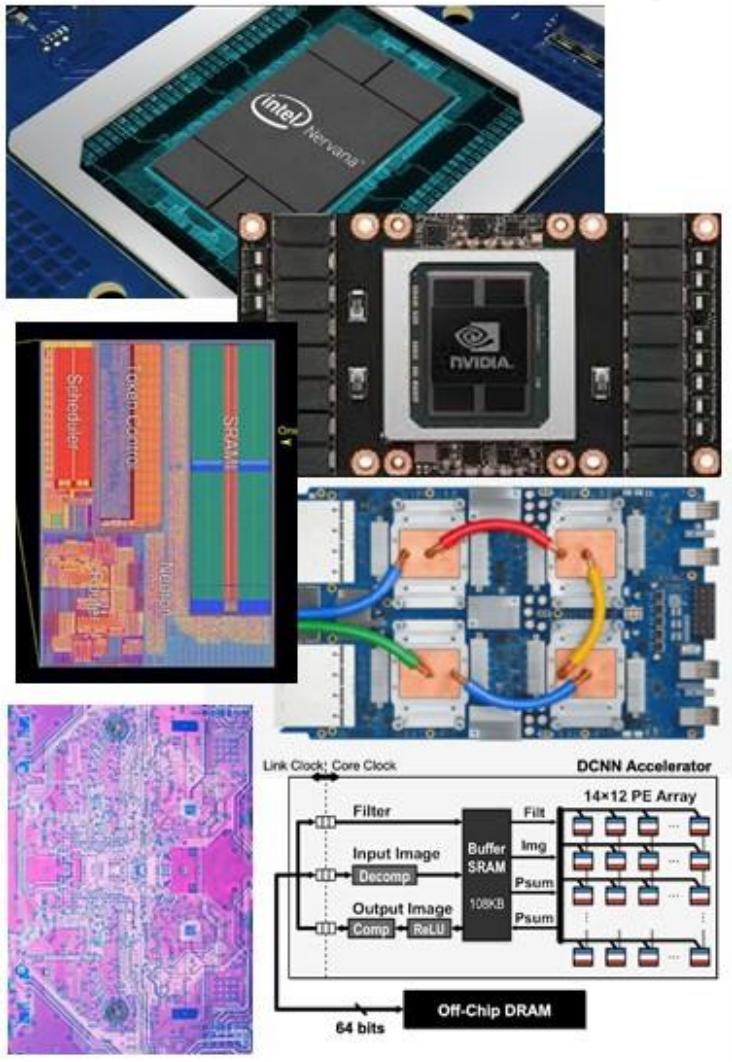
Embedded Devices

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Apa itu embedded  
machine learning?



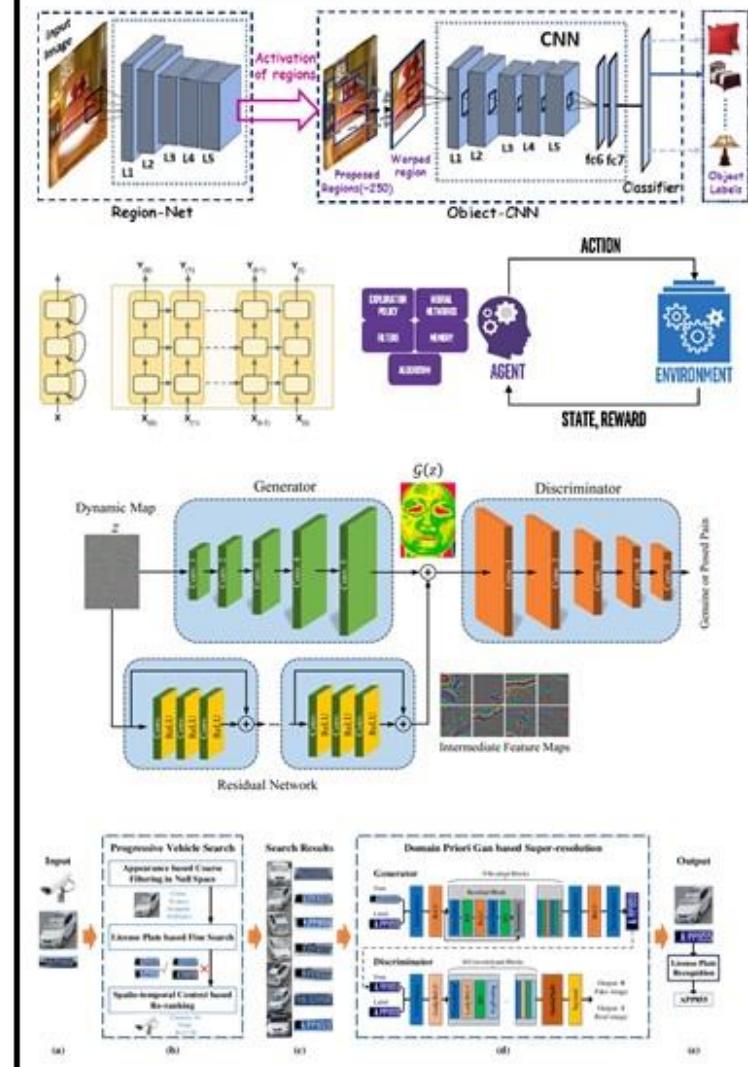
## Hardware Accelerators



## Embedded and IoT Systems



## Deep Learning Software



Referensi:

<https://www.engr.colostate.edu/~sudeep/embedded-machine-learning/>

# Keuntungan ML pada embedded device

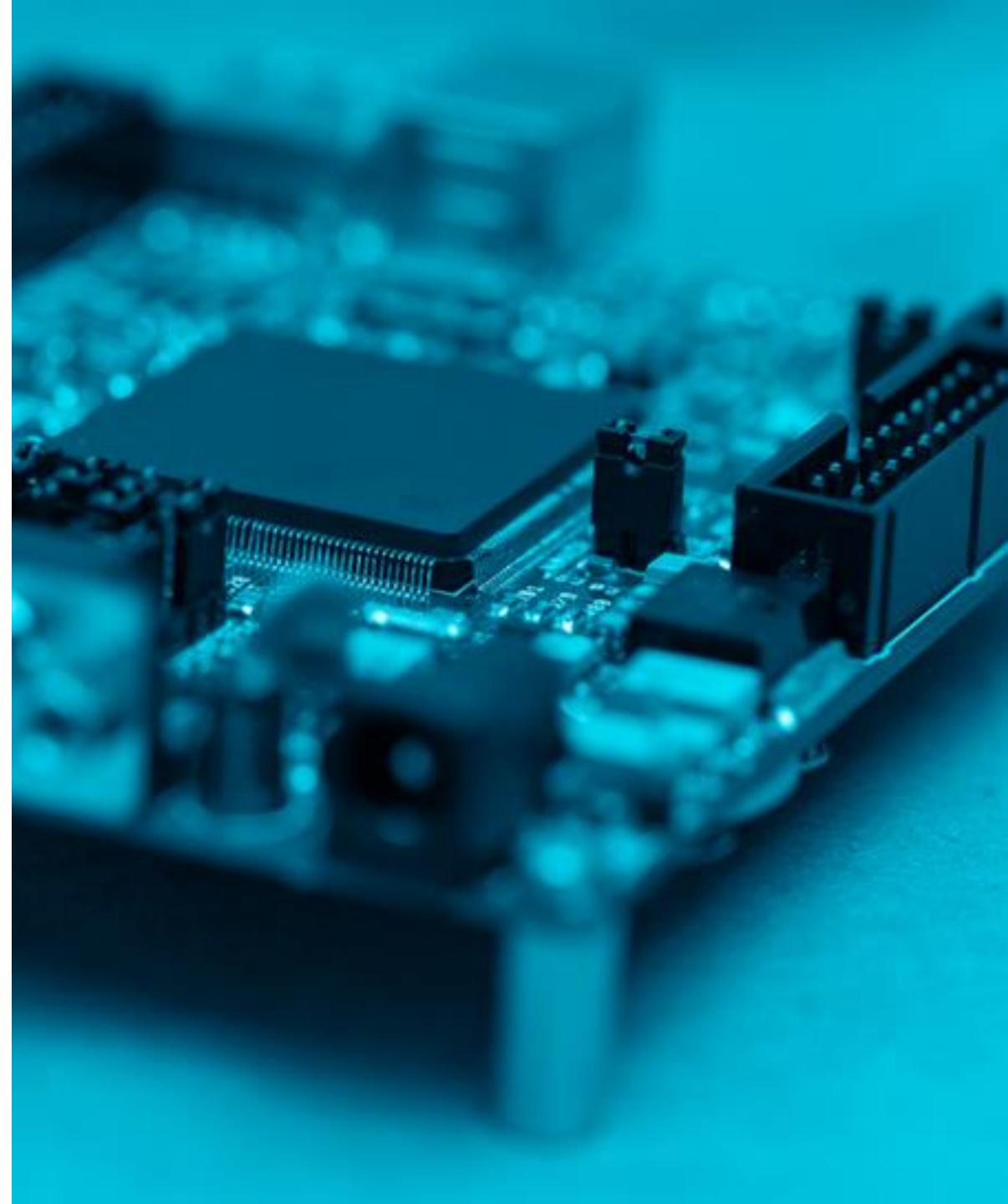
- **Bandwidth**—ML algorithms on edge devices can extract meaningful information from data that would otherwise be inaccessible due to bandwidth constraints.
- **Latency**—On-device ML models can respond in real-time to inputs, enabling applications such as autonomous vehicles, which would not be viable if dependent on network latency.
- **Economics**—By processing data on-device, embedded ML systems avoid the costs of transmitting data over a network and processing it in the cloud.
- **Reliability**—Systems controlled by on-device models are inherently more reliable than those which depend on a connection to the cloud.
- **Privacy**—When data is processed on an embedded system and is never transmitted to the cloud, user privacy is protected and there is less chance of abuse.

*Referensi:*

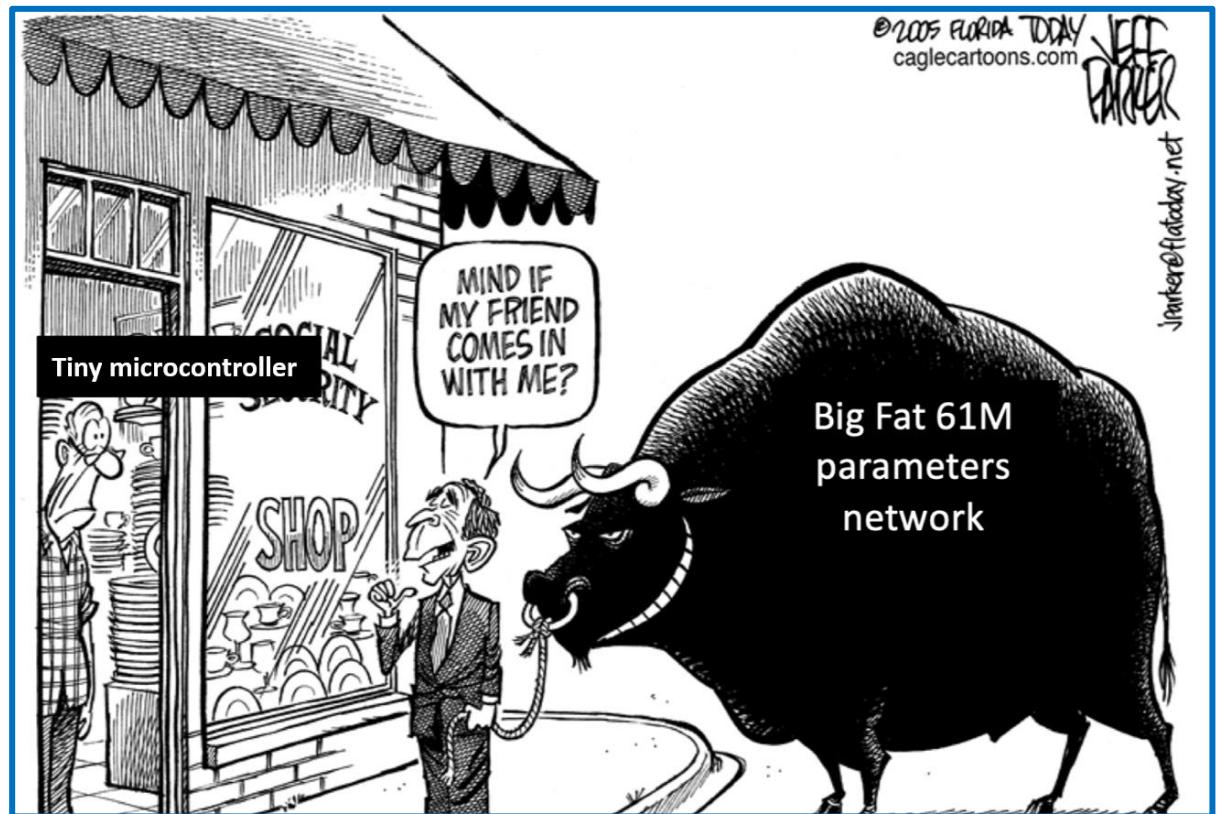
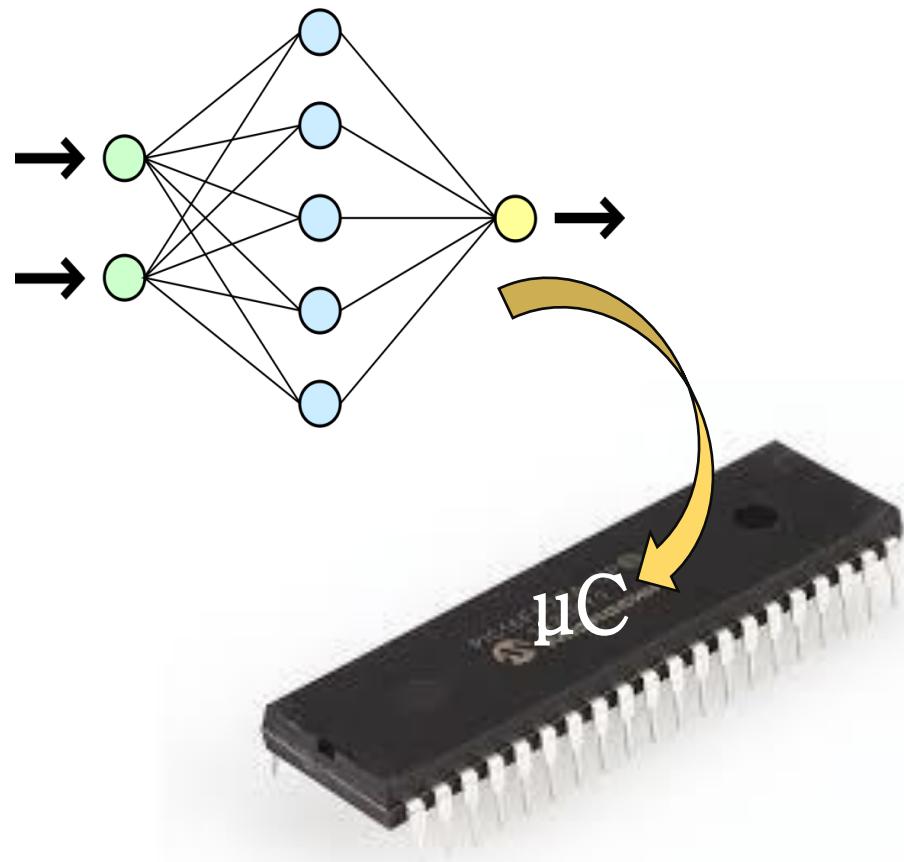
<https://docs.edgeimpulse.com/docs/what-is-embedded-machine-learning-anyway>

---

Pernah mendengar  
tinyML?

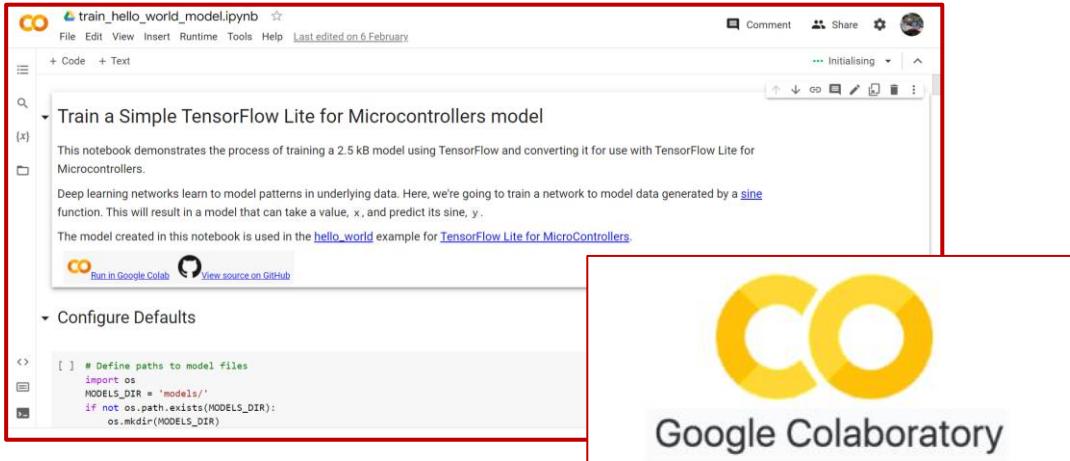


# TinyML



## Machine Learning:

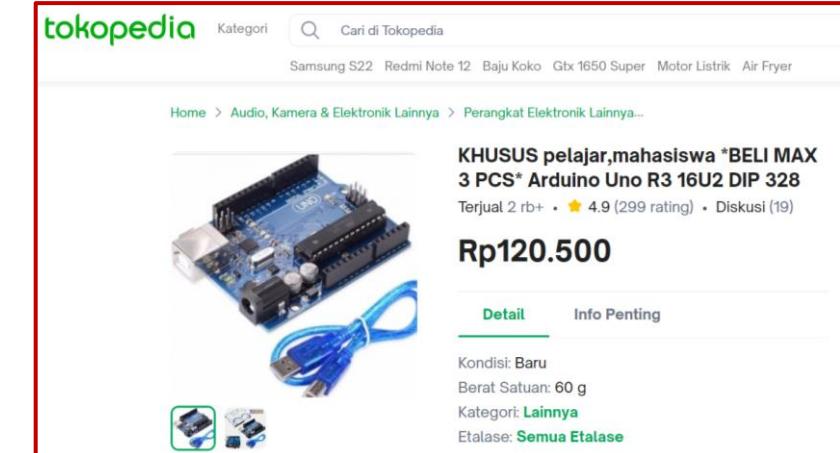
- ❑ Arsitektur neural network semakin canggih
  - Gratis, open-source, framework mudah digunakan
- ❑ End-to-end Tools → membuat desain ML semakin sederhana
- ❑ ML tidak hanya untuk para ahli!



The screenshot shows a Google Colaboratory notebook titled "train\_hello\_world\_model.ipynb". The notebook content includes a section titled "Train a Simple TensorFlow Lite for Microcontrollers model" which demonstrates the process of training a 2.5 kB model using TensorFlow and converting it for use with TensorFlow Lite for Microcontrollers. It also includes a "Configure Defaults" section with some Python code. A red box highlights the "Google Colaboratory" logo at the bottom right of the interface.

## Mikrokontroler:

- ❑ Dapat diprogram dengan mudah
  - Lingkungan pengembangan (IDE) yang menawarkan plug-and-play
  - Dengan sedikit baris koding kita dapat membaca sensor, menyalakan LED, memutat motor, dsb
  - Koding kompatibel dgn berbagai platform
- ❑ Terdapat banyak sensor dan papan pengembangan mikrokontroler dengan biaya terjangkau



The screenshot shows a product listing on Tokopedia for an "Arduino Uno R3 16U2 DIP 328". The product is highlighted with a red box. The listing includes a thumbnail image of the blue microcontroller board with various components attached, a price of "Rp120.500", and some product details like "Kondisi: Baru", "Berat Satuan: 60 g", "Kategori: Lainnya", and "Etalase: Semua Etalase".

# Penerapan TinyML: 3V



## Vibration

- Accelerometer
- Suhu, kelembapan, dsb



## Vision

- Image
- Video



## Voice

- Pengenalan suara/bunyi
- NLP

# Perusahaan besar di balik pengembangan tinyML

AONdevices

arm

Deeplite

EDGE IMPULSE

emza  
visual sense

FotaHub

GREENWAVES  
TECHNOLOGIES



Grovety Inc.

HOTG

imagimob

itemis

KLIKA·TECH  
GLOBAL IOT SOLUTIONS

LatentAI  
Adaptive AI for a Smarter Edge

maxim integrated.  
NOW PART OF  
ANALOG  
DEVICES

Micro.ai

NXP

PROPHESEE  
METAVISION FOR MACHINES

Qualcomm

Qeexo

RealityAI

RENESAS

REEXEN  
technology

SAP

seeed  
STUDIO  
The IoT Hardware Enabler

SensiML™

Sony  
Semiconductor  
Solutions  
Corporation

ST  
life.augmented

STREAM ANALYZE



SynSense

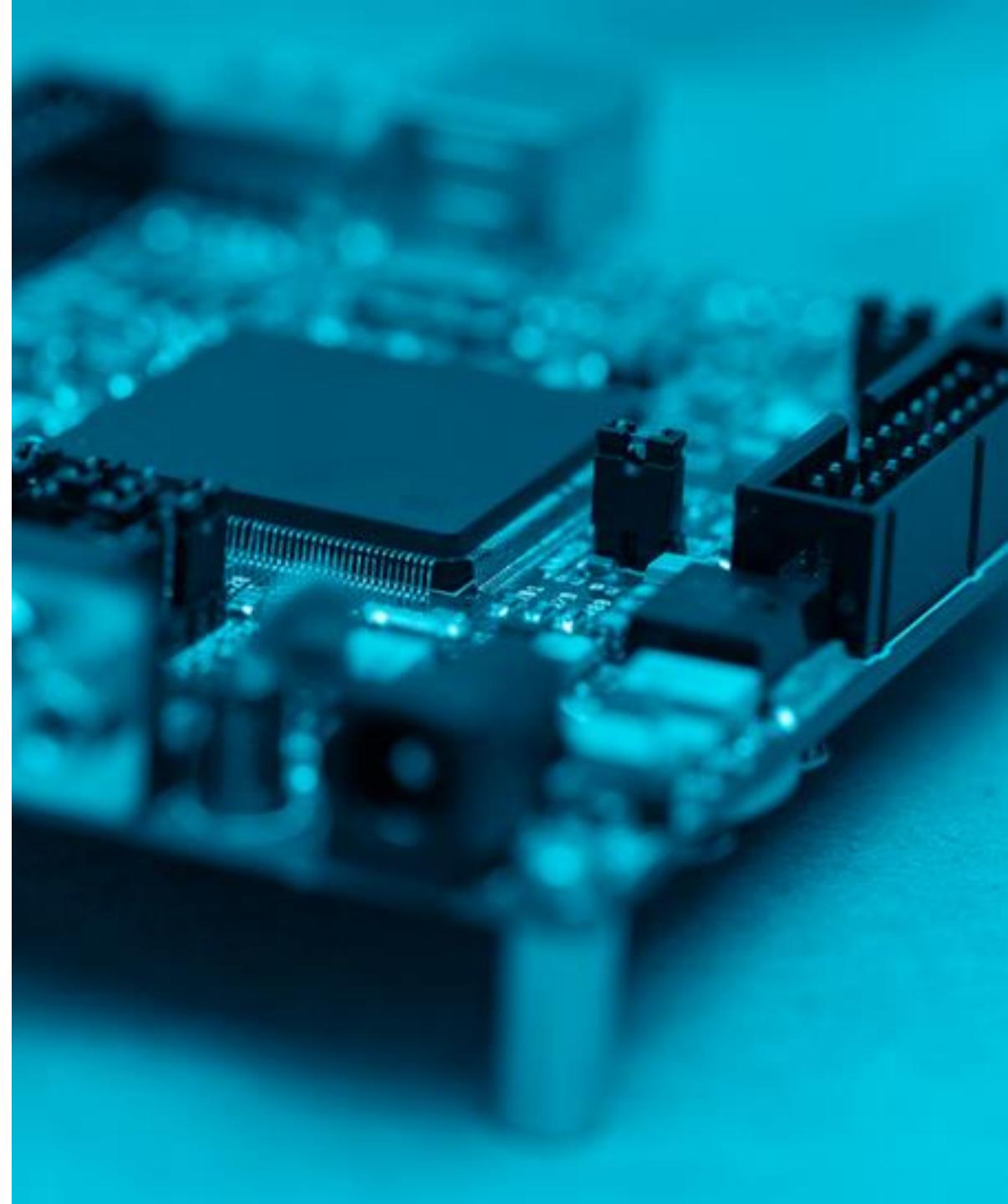
SYNTIANT

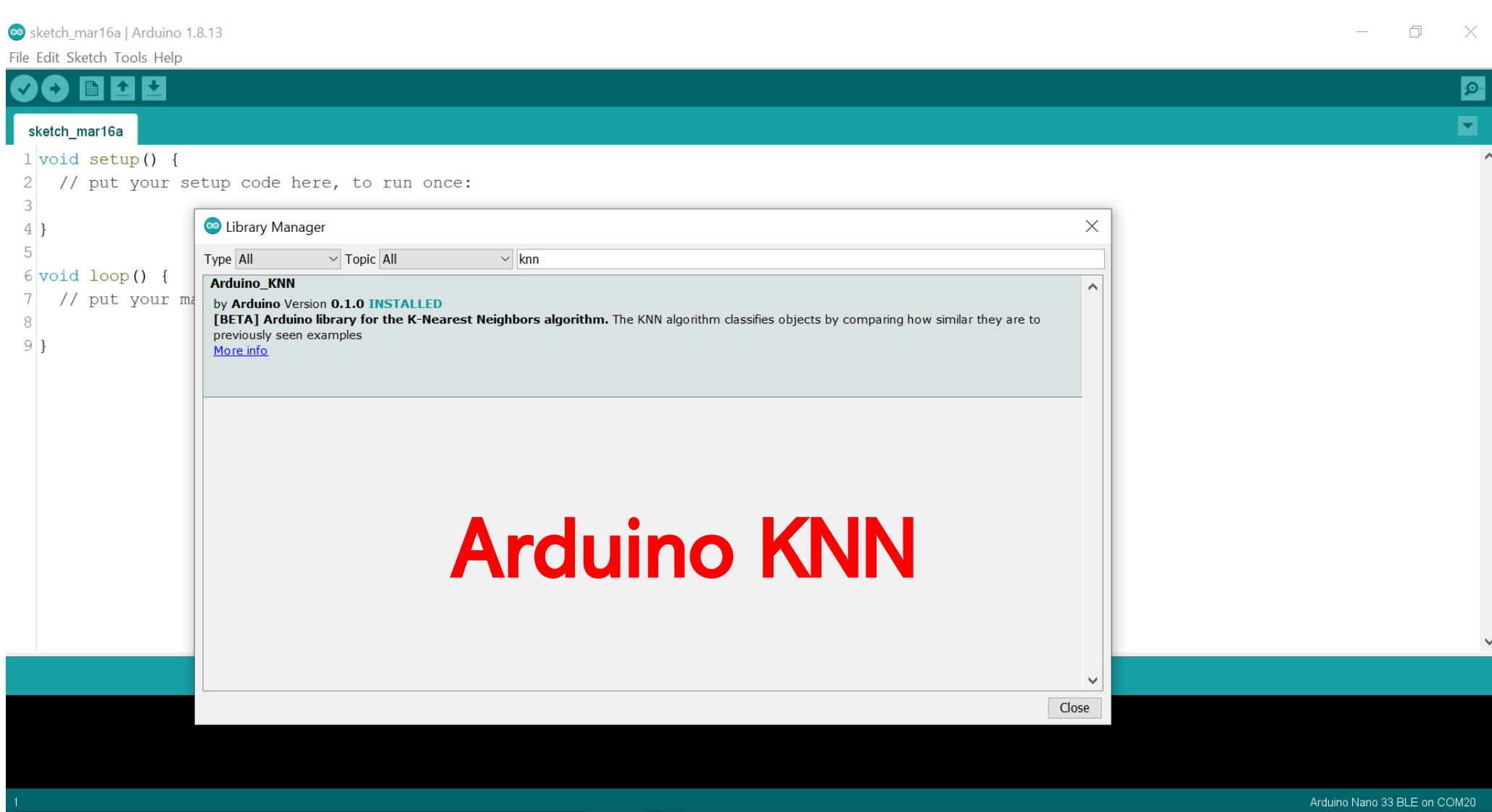
Referensi:

<https://www.tinyml.org/>

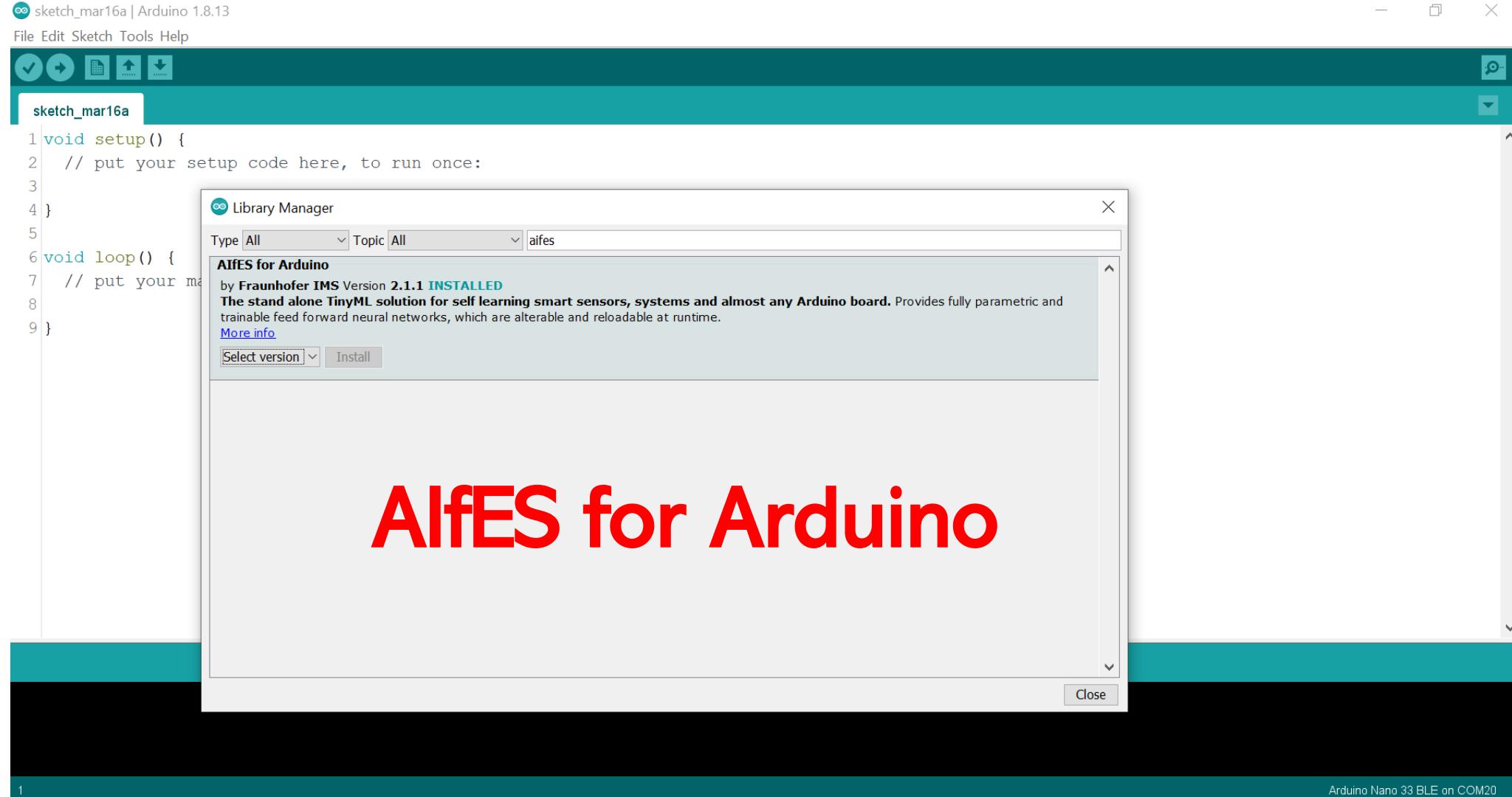
---

Oke, saya tertarik (sedikit) tentang tinyML. Dari mana saya memulai?





<https://blog.arduino.cc/2020/06/18/simple-machine-learning-with-arduino-knn/>



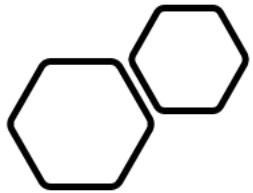
# AlfES for Arduino

AlfES (Artificial Intelligence for Embedded Systems) → [https://github.com/Fraunhofer-IMS/AlfES\\_for\\_Arduino](https://github.com/Fraunhofer-IMS/AlfES_for_Arduino)

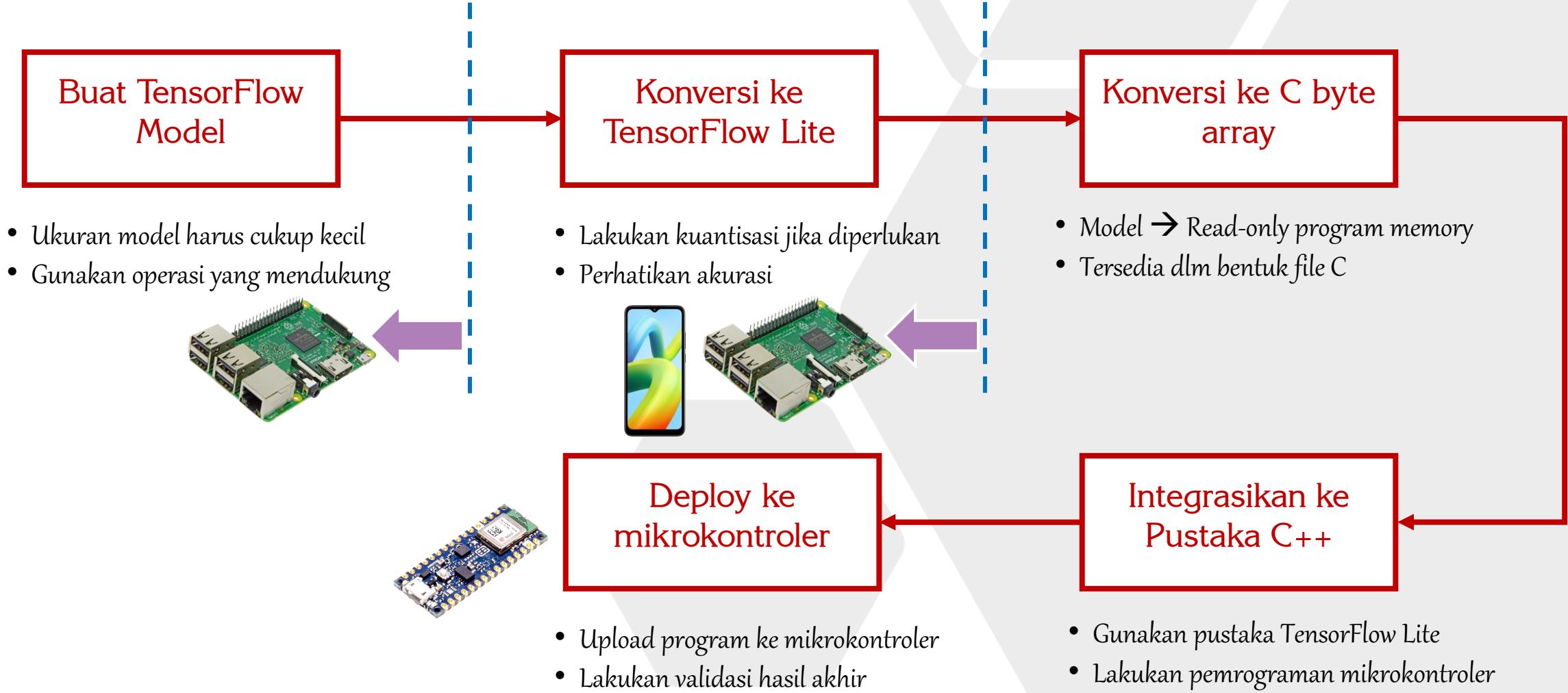
# TensorFlow Lite

- **TF Lite** → open-source deep learning framework untuk menjalankan model TensorFlow pada alat/device
- **Target** : mobile, embedded & IoT devices
- Memungkinkan menjalankan **on-device machine learning** → low latency & small binary size.
- **Penerapan**: Mulai dari mikrokontroler sampai pada perangkat bergerak



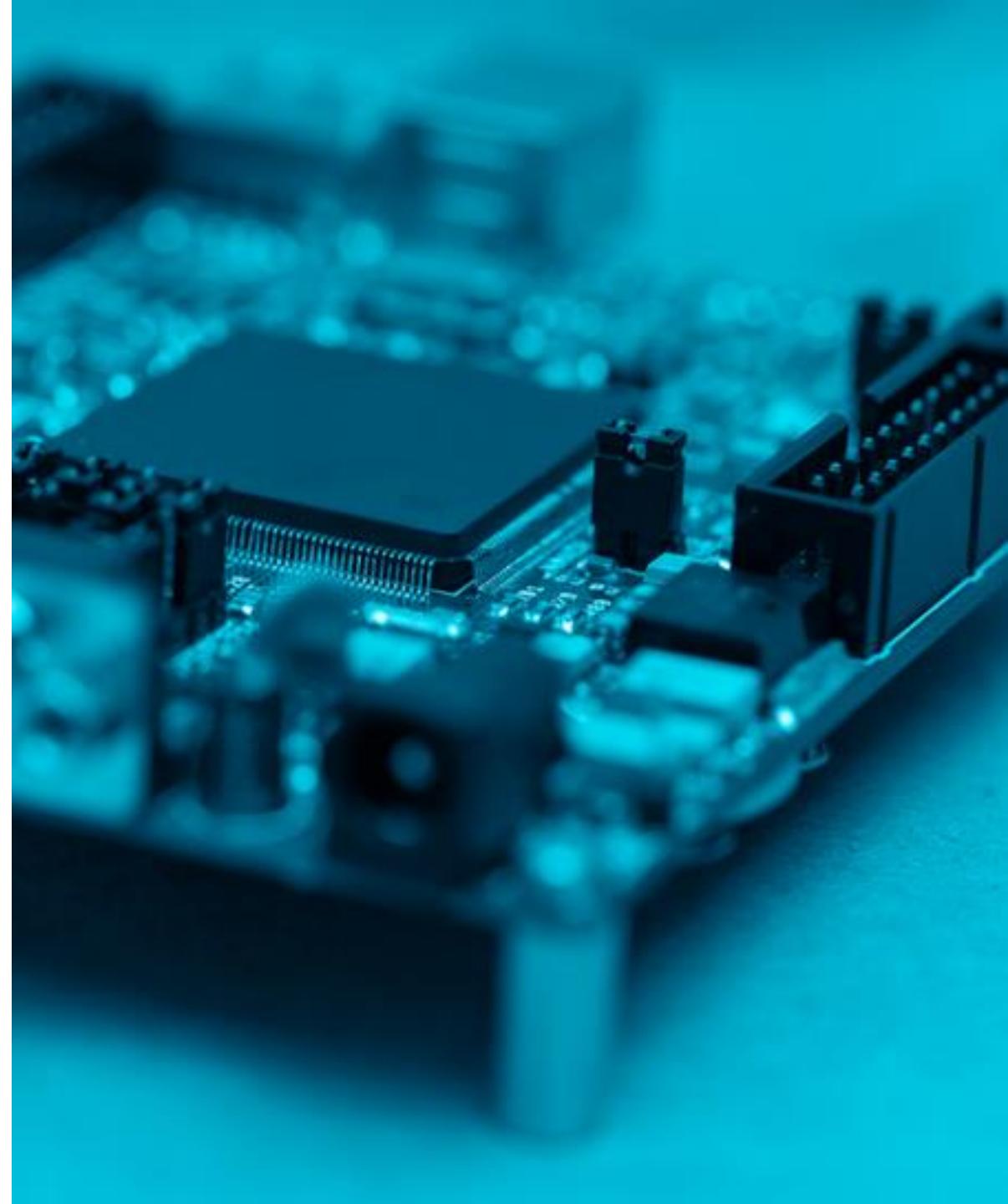


# TF Lite Developer workflow

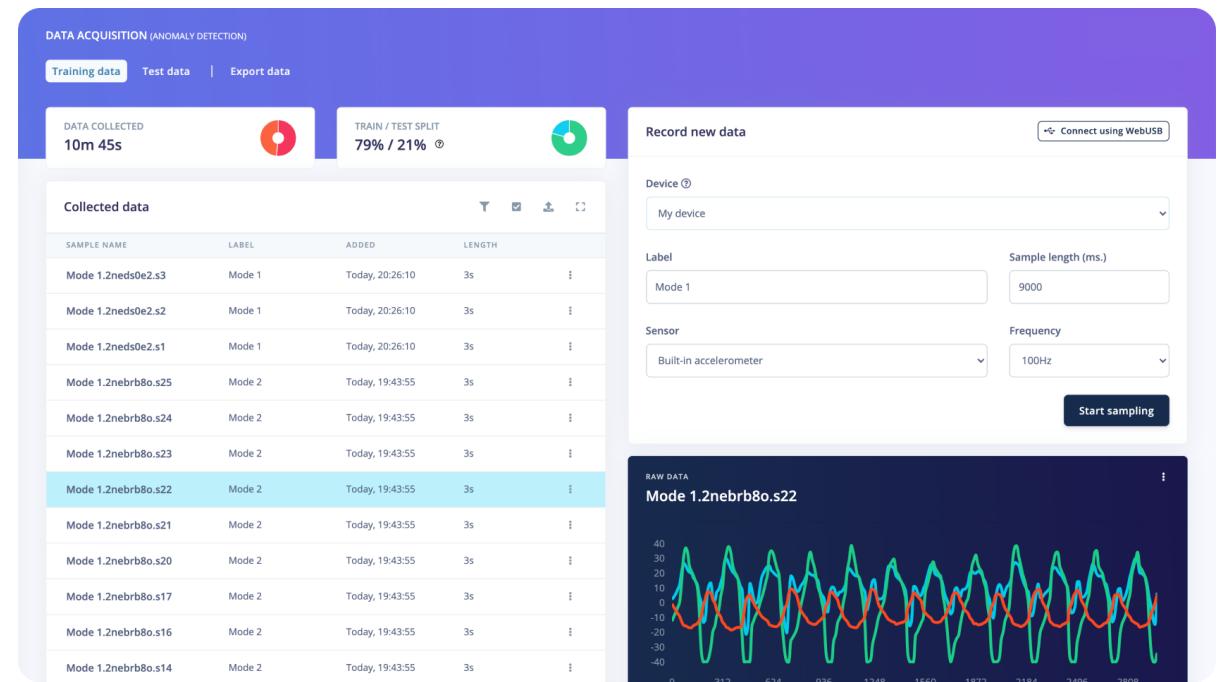
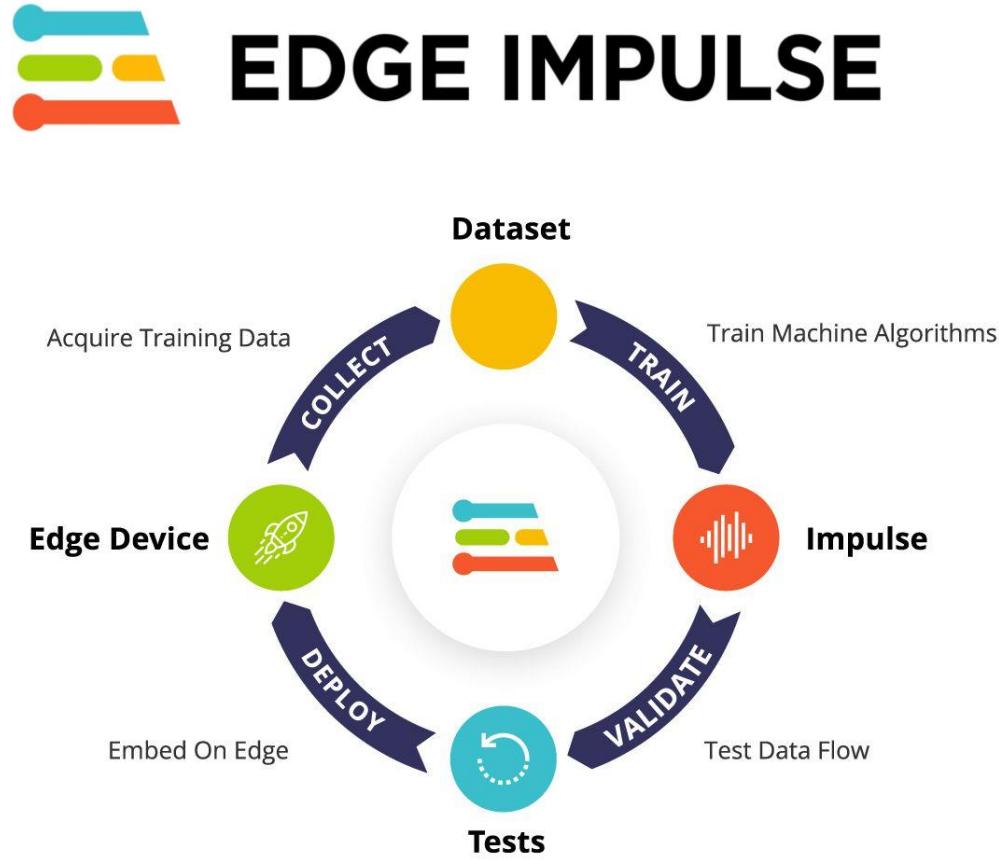


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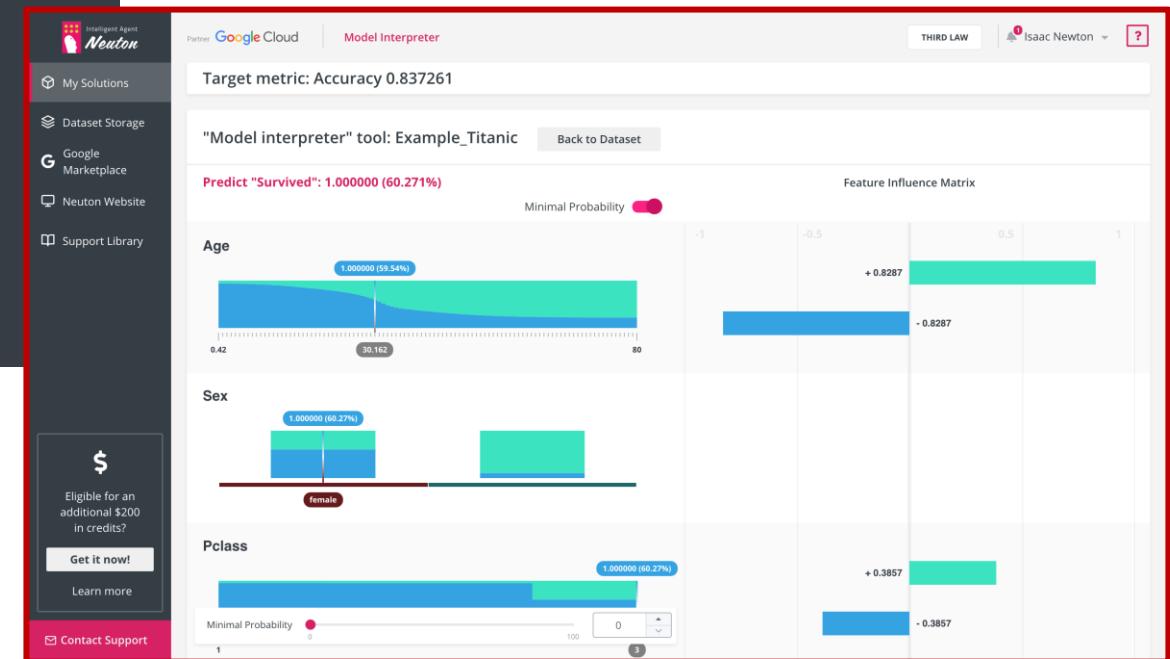
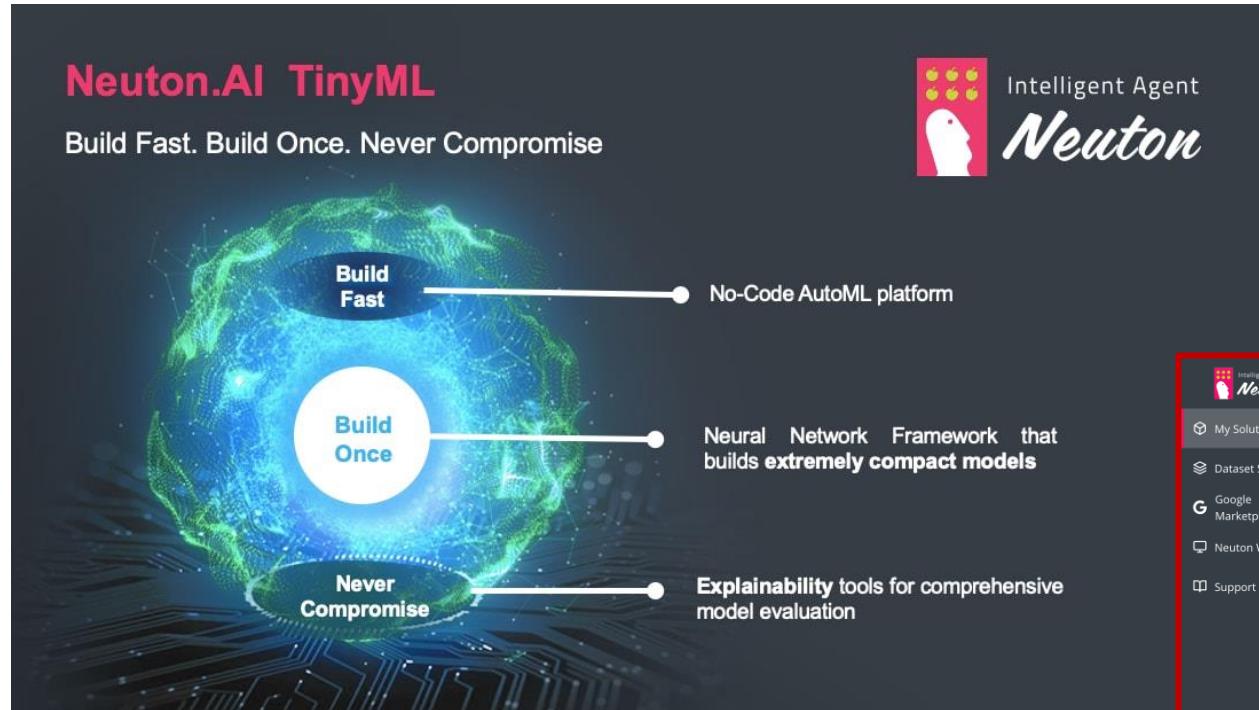
Saya tidak ingin ribet. Bisa kasih saya platform yang lebih sederhana?



# Gunakan Auto(tiny)ML

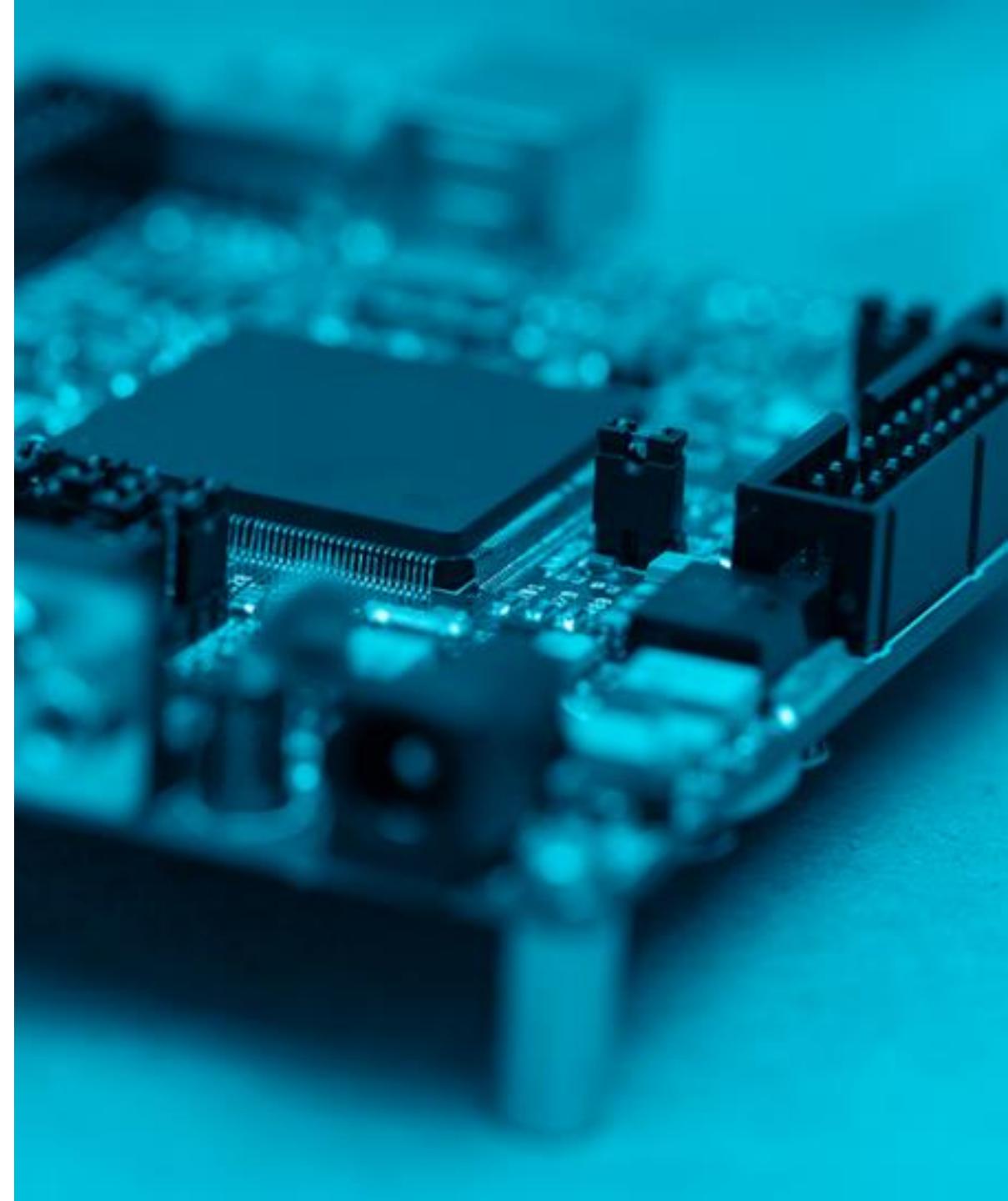


# Gunakan Auto(tiny)ML

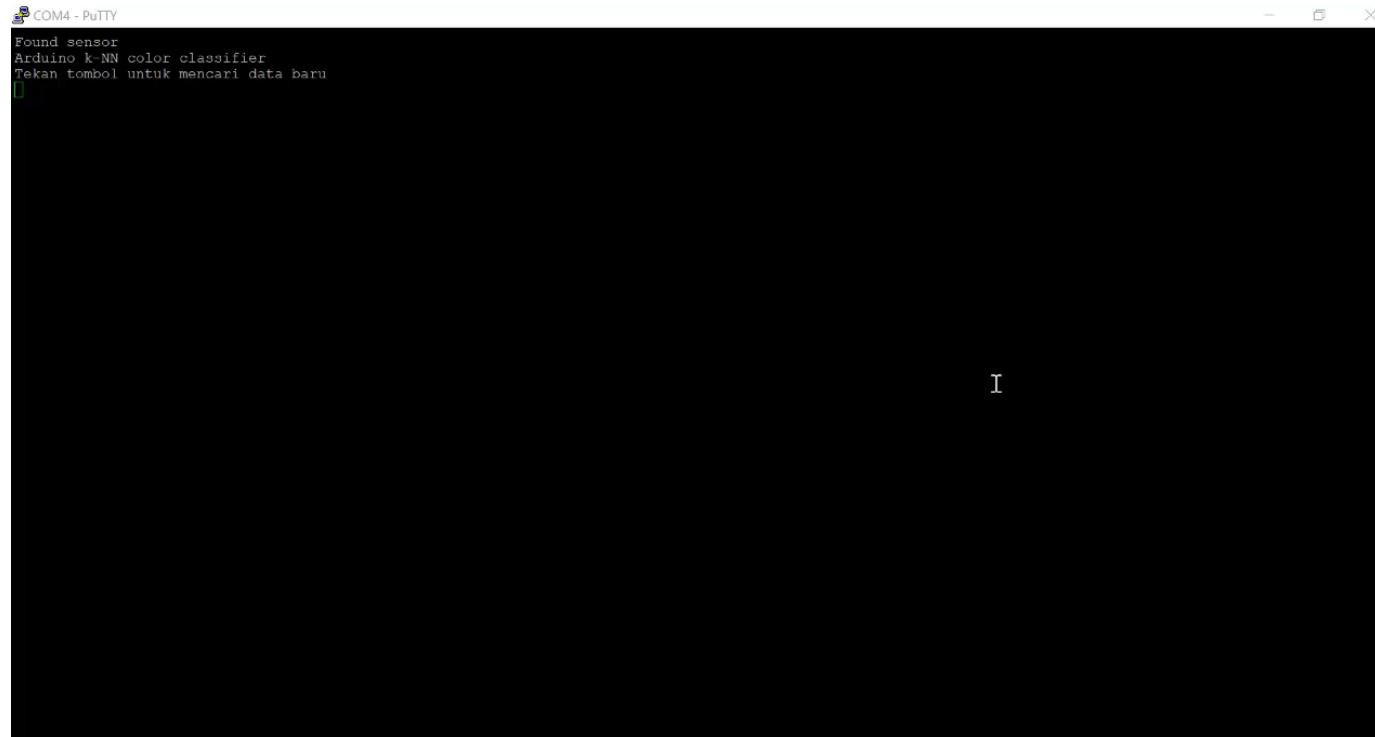


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Bisa kasih saya sedikit  
demo mengenai  
penerapan AI?



# Penerapan TinyML Klasifikasi Buah Menggunakan KNN - Arduino



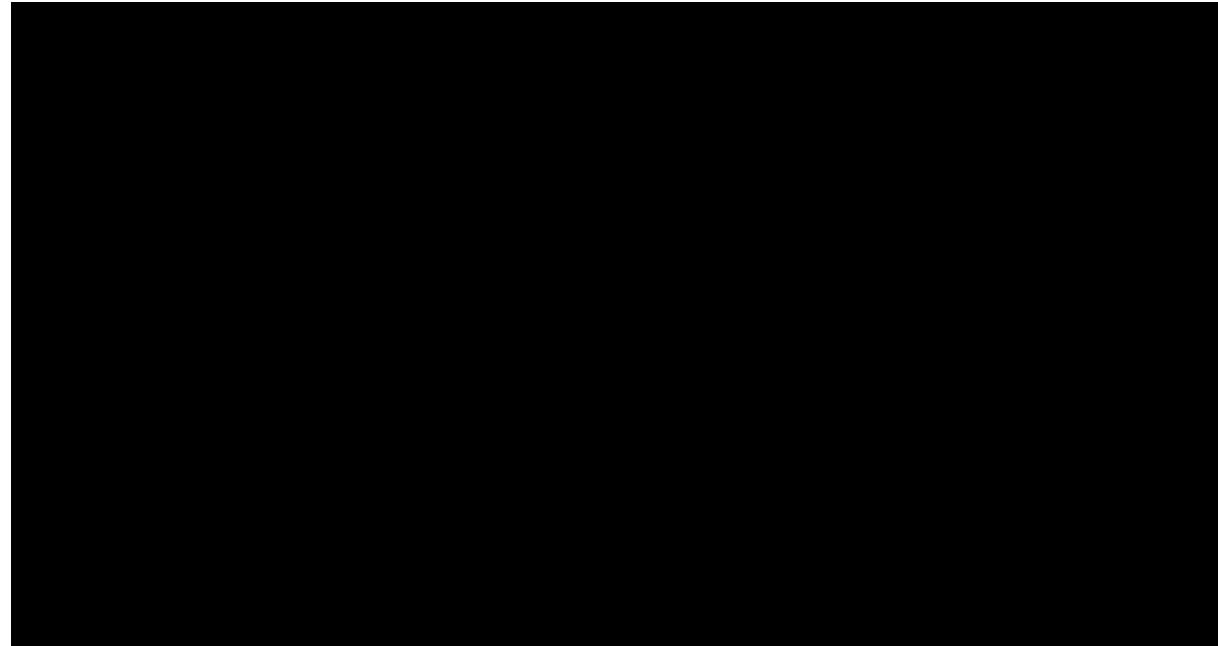
Video Link:

[https://github.com/Narin-Laboratory/SeminarUndiknas/blob/main/colour\\_classification\\_fruits.mp4](https://github.com/Narin-Laboratory/SeminarUndiknas/blob/main/colour_classification_fruits.mp4)

# Penerapan TinyML

## Motion Recognition Menggunakan Edge Impulse

- Pengambilan Data

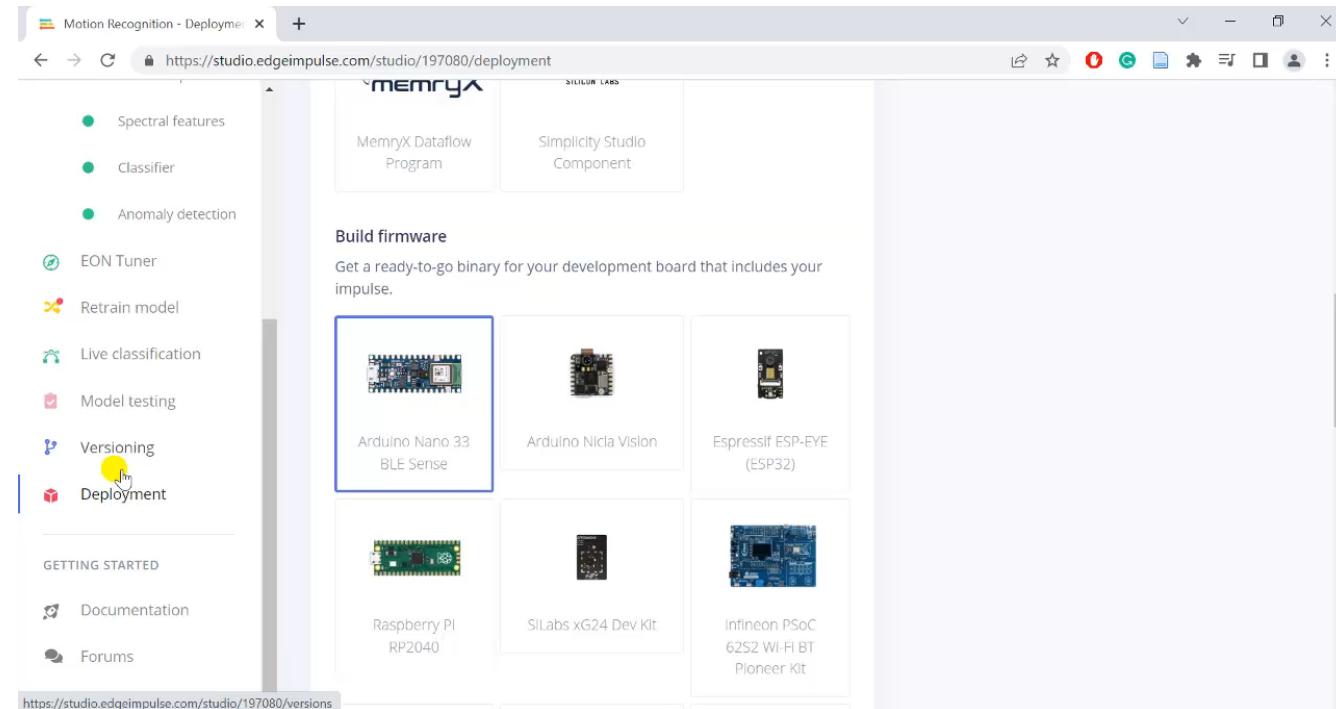


Video Link:

[https://github.com/Narin-Laboratory/SeminarUndiknas/blob/main/ambil\\_data\\_gesture.mp4](https://github.com/Narin-Laboratory/SeminarUndiknas/blob/main/ambil_data_gesture.mp4)

# Penerapan TinyML: Edge Impulse

- Deployment ke Alat



Video Link:

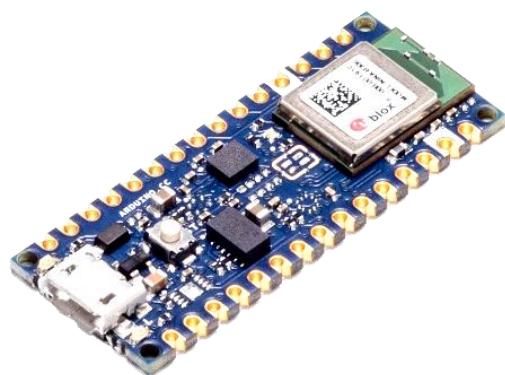
<https://github.com/Narin-Laboratory/SeminarUndiknas/blob/main/deployment.mp4>

# Contoh menggunakan TensorFlow: Pengenalan Hand Gesture Sederhana

- Demo resources:

<https://github.com/Narin-Laboratory/Webinar/tree/master/HandGesture>

## Hardware



## Software



TensorFlow

TensorFlow Lite



ARDUINO

matplotlib

pandas

NumPy

# Definisi Project

- Membedakan dua jenis gesture sederhana:

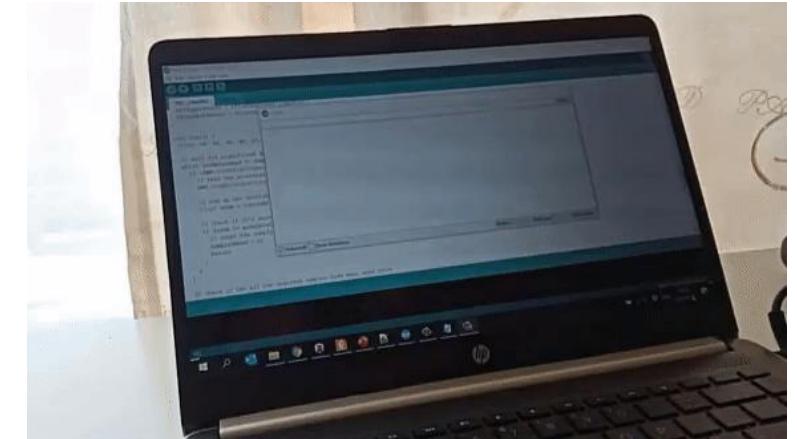
Kepalan tangan **telungkup**

Kepalan tangan **tengadah**

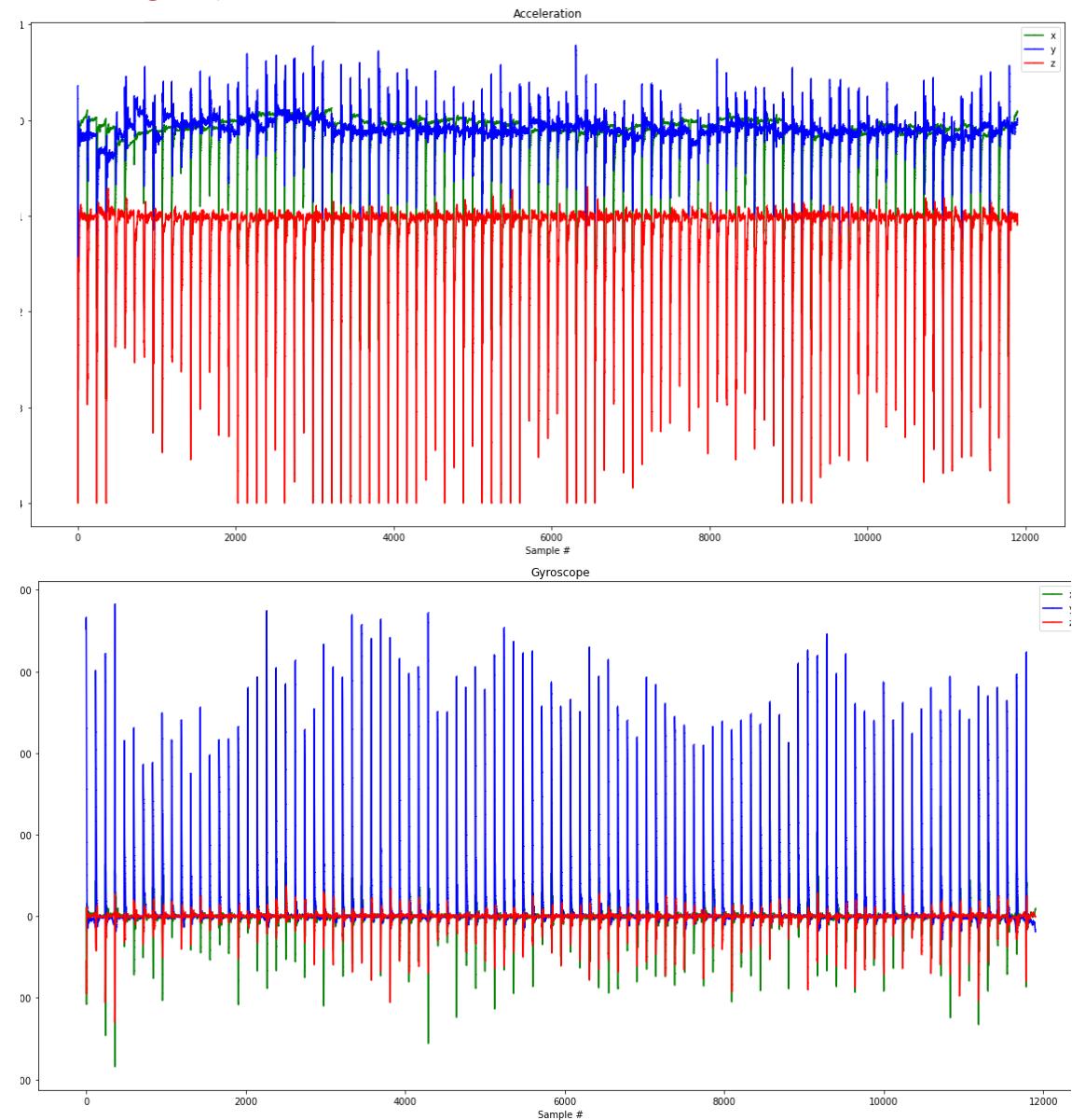
- Data:

Output dari sensor IMU → accelerometer & gyroscope

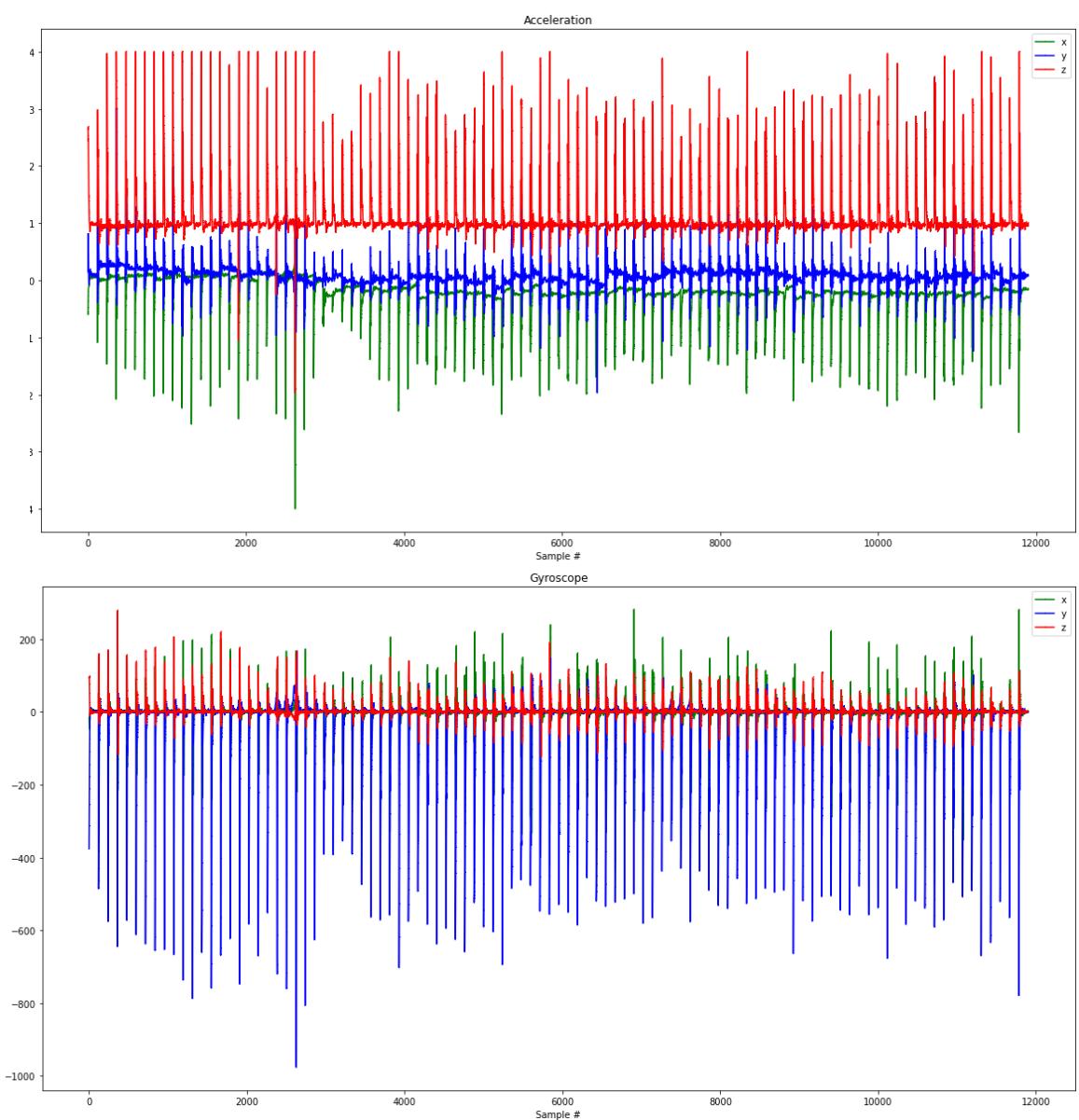
Model arsitektur: Fully connected layer



## Telungkup

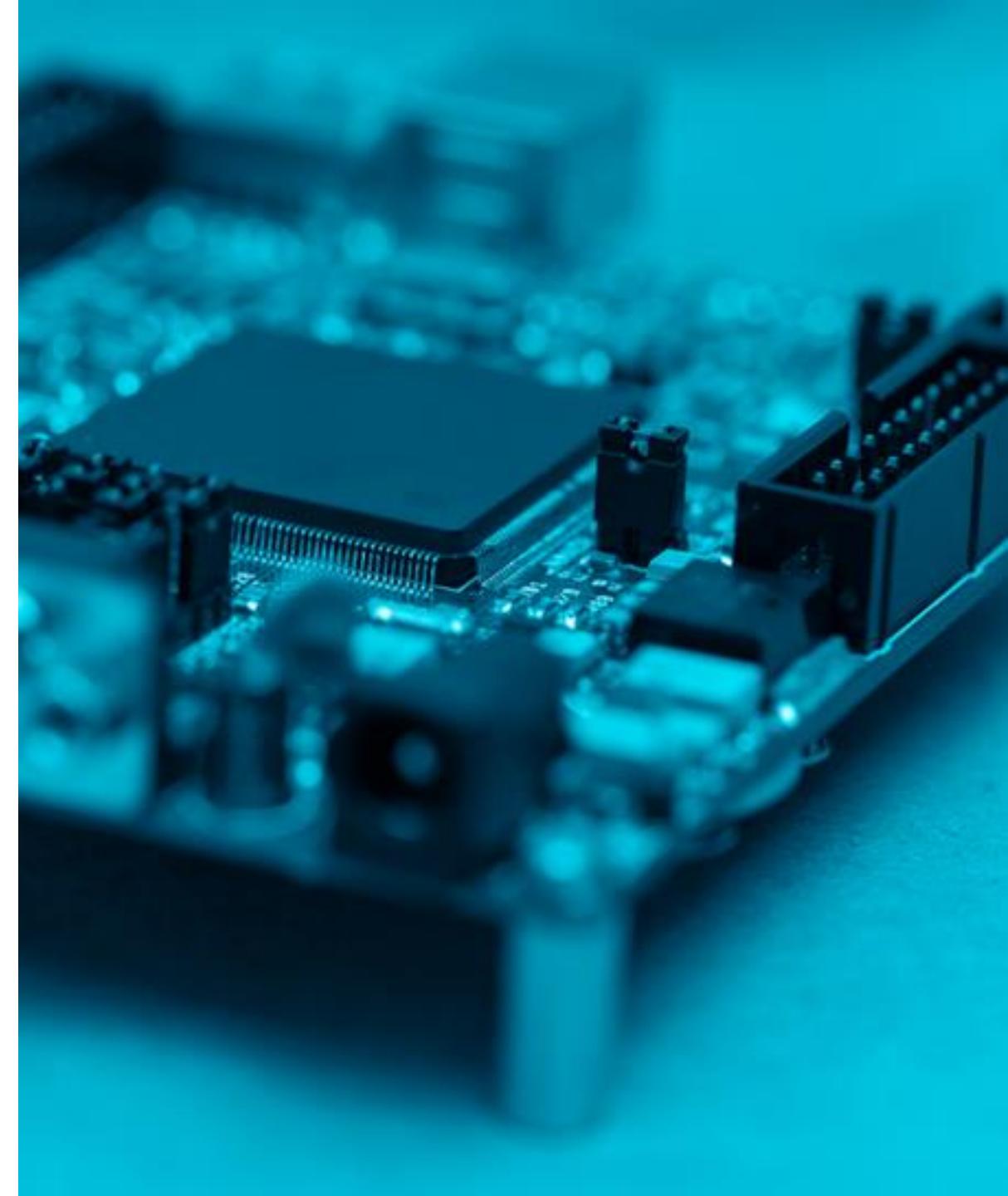


## Tengadah



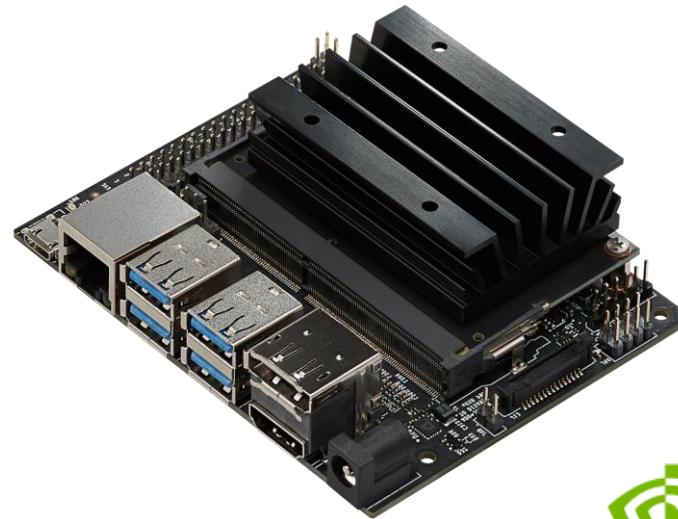
---

Saya tidak mau menggunakan mikrokontroler, tapi saya suka koding.  
Ada saran agar bisa menggunakan ML untuk bidang embedded?



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# Gunakan Raspberry Pi atau SBC yang lain



# Contoh: Prediksi Energi Listrik di Rumah Hunian

- Demo resources:

<https://github.com/Narin-Laboratory/Webinar/tree/master/EnergyPrediction>

- Reference resources:

<https://github.com/LuisM78/Appliances-energy-prediction-data>

<https://archive.ics.uci.edu/ml/datasets/Appliances+energy+prediction>

## Hardware



Raspberry Pi3

## Software



TensorFlow



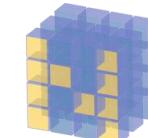
scikit  
learn

matplotlib



TensorFlow Lite

pandas



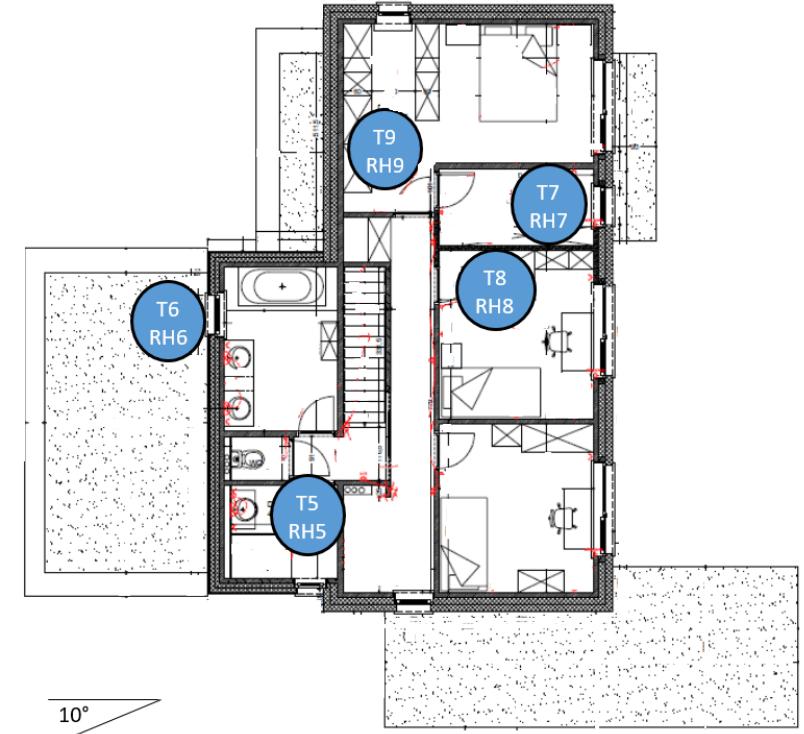
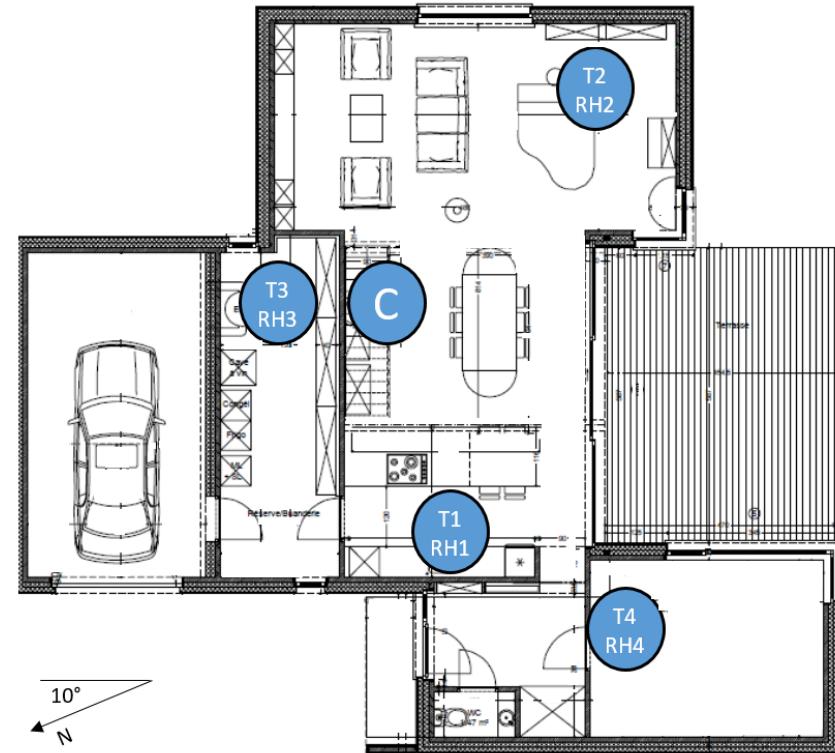
NumPy

Seaborn

# Contoh Penerapan: Prediksi Energi Listri di Rumah Hunian



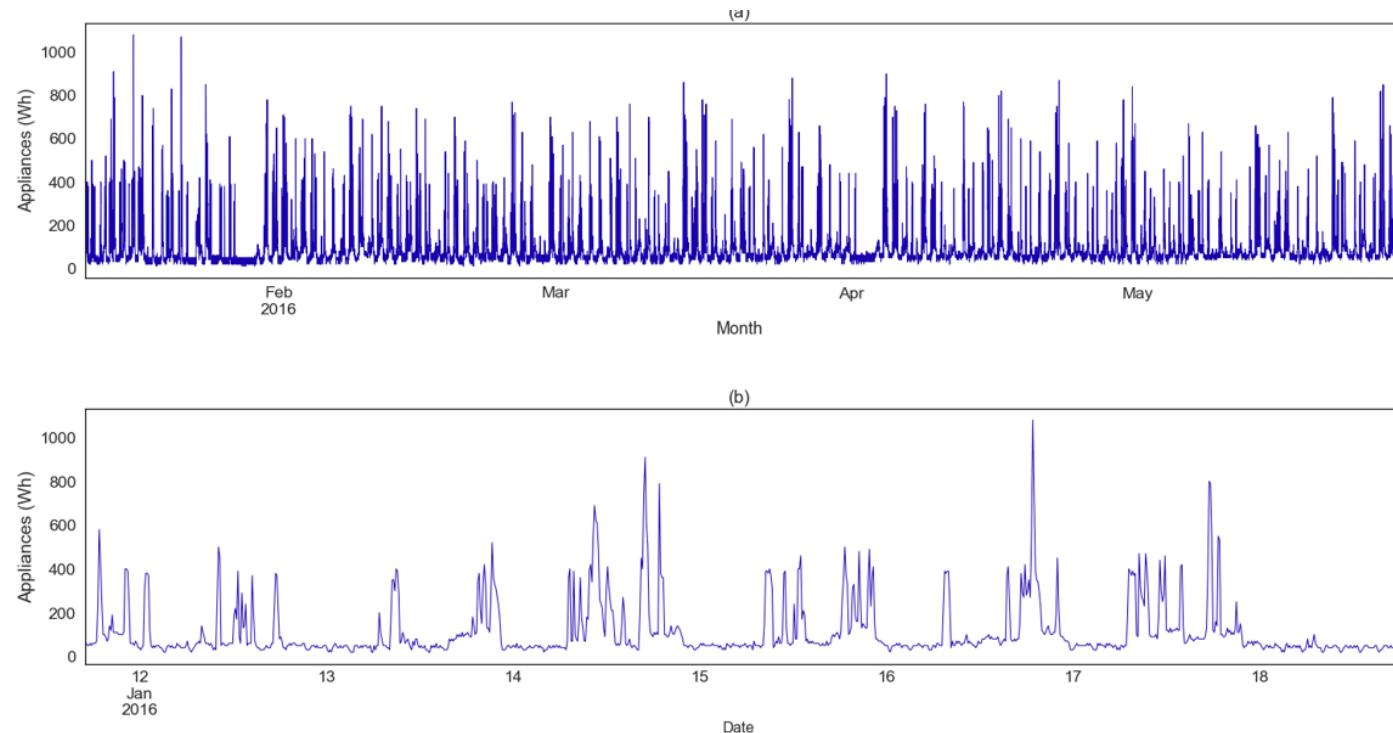
Data Driven Prediction Model



L. M. Candanedo, V. Feldheim, and D. Deramaix, "Data driven prediction models of energy use of appliances in a low-energy house," Energy Build., vol. 140, pp. 81–97, Apr. 2017, doi: 10.1016/j.enbuild.2017.01.083.

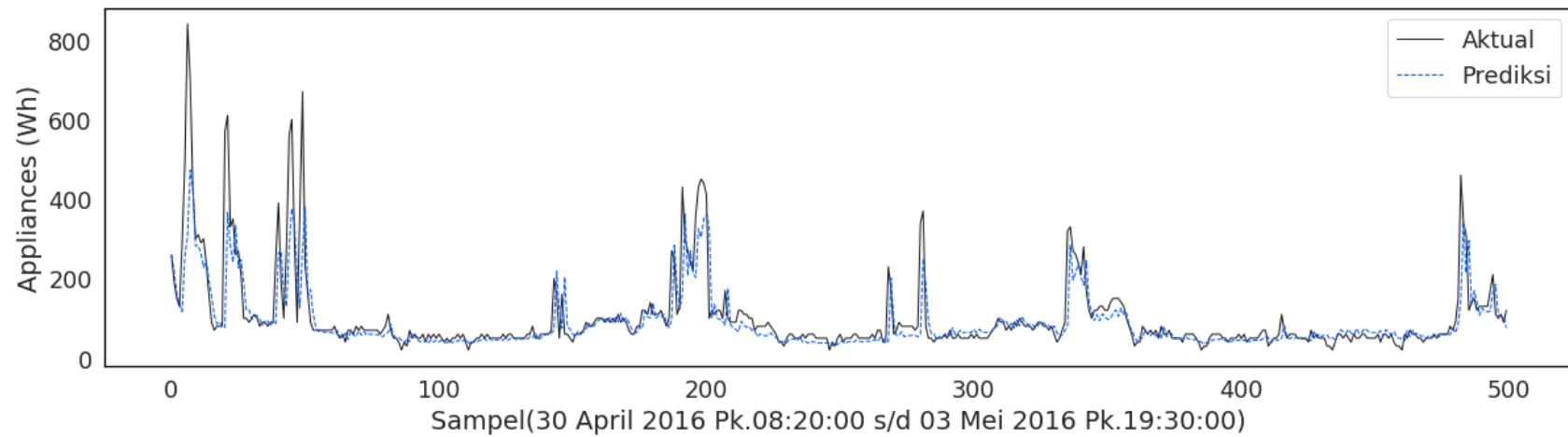
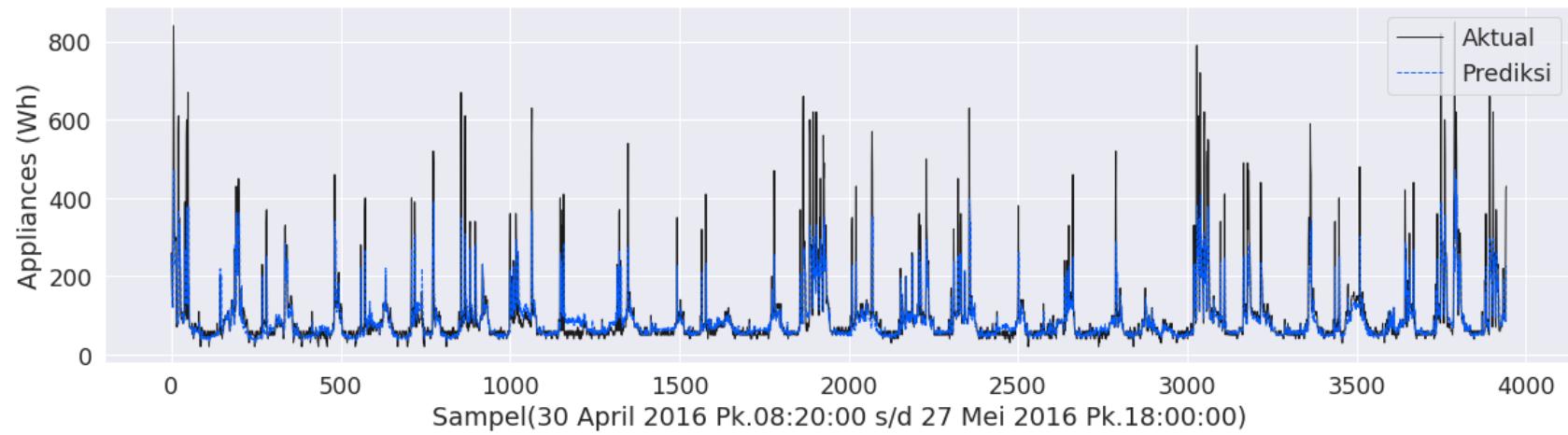
# Studi Kasus 2: Prediksi Energi Listrik di Rumah Hunian

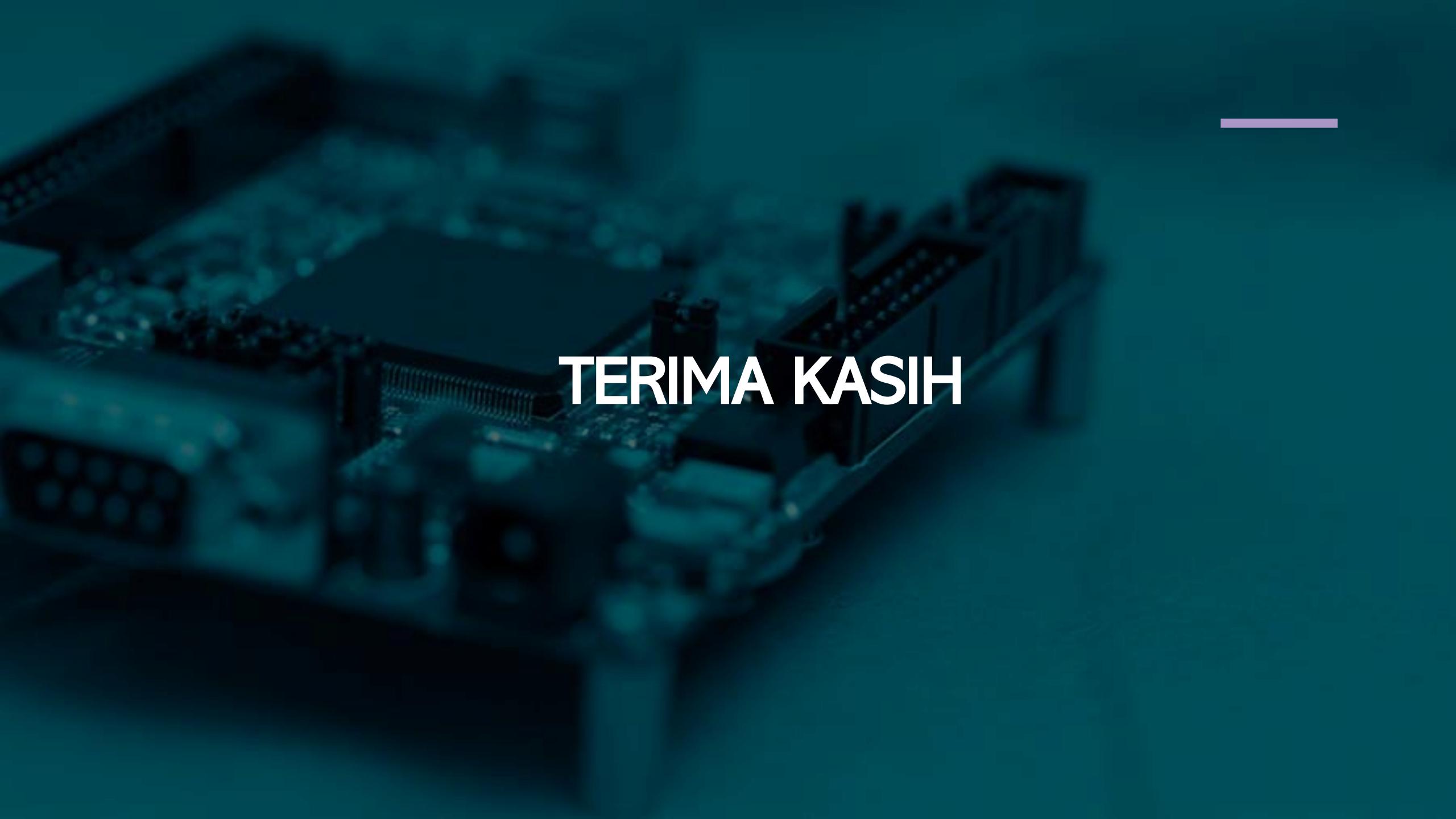
- Tipe: Regresi
- Model arsitektur: LSTM
- Input: lighting, suhu, kelembaban, kecepatan angin, dsb
- Target: memprediksi kebutuhan energi listrik
- Jumlah data: 19.735 baris data
- Pembagian data:
  - ✓ 60% sebagai data latih (11.841 baris)
  - ✓ 20% sebagai data validasi (3.947 baris)
  - ✓ 20% sebagai data uji (3.947 baris)



Gambar 1. (a) Pola konsumsi energi listrik selama periode penuh, (b) Tinjauan lebih detail untuk pola konsumsi energi selama satu minggu pertama

# Hasil Prediksi





TERIMA KASIH

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