

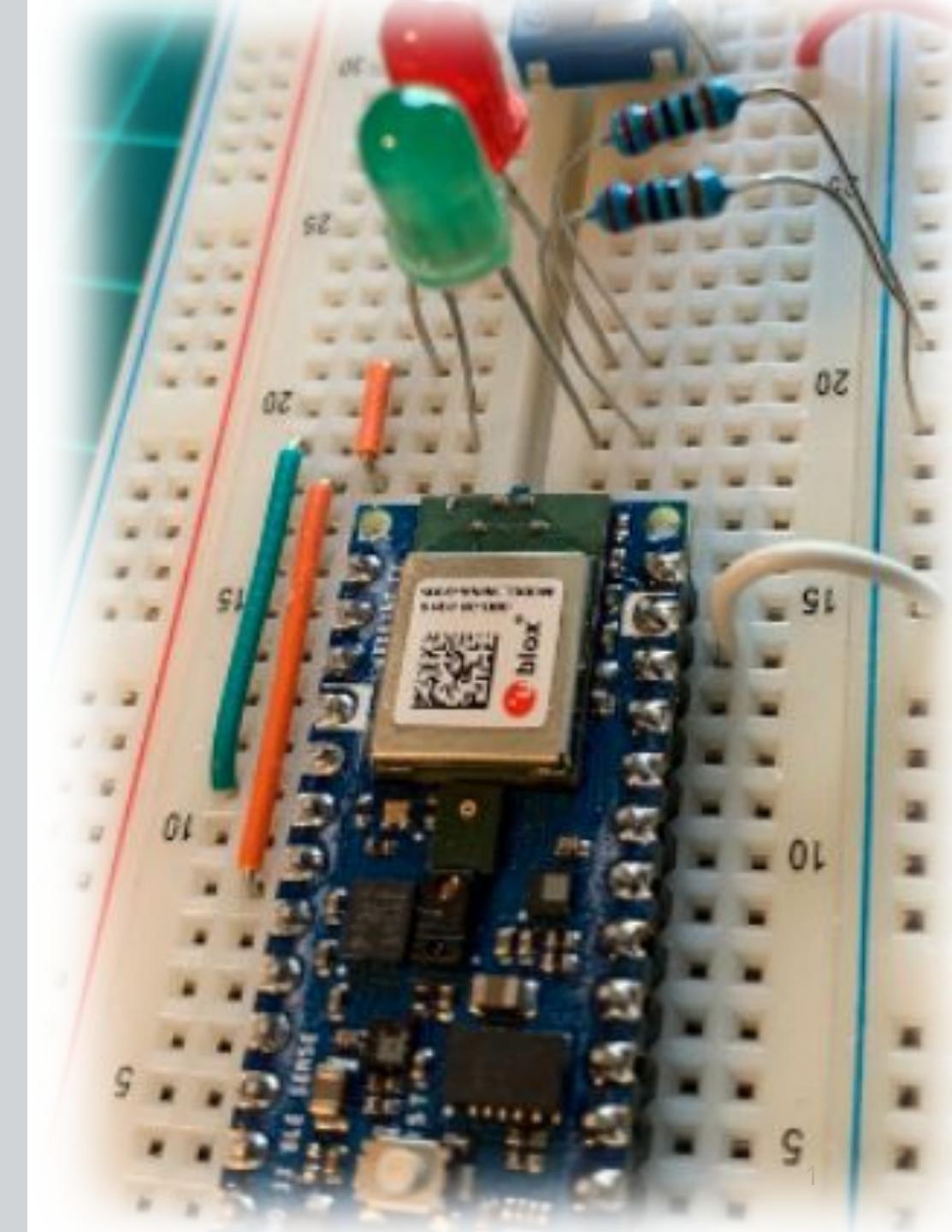
IESTI01 – TinyML

Embedded Machine Learning

15. ML Applications Overview AI Lifecycle and ML Workflow



Prof. Marcelo Rovai
UNIFEI



TinyML Application

Examples (Classification)

Predictive Maintenance



Motion, current, audio and camera

- Industrial
- White goods
- Infrastructure
- Automotive

Asset Tracking & Monitoring



Motion, temp, humidity, position, audio and camera

- Logistics
- Infrastructure
- Buildings
- Agriculture

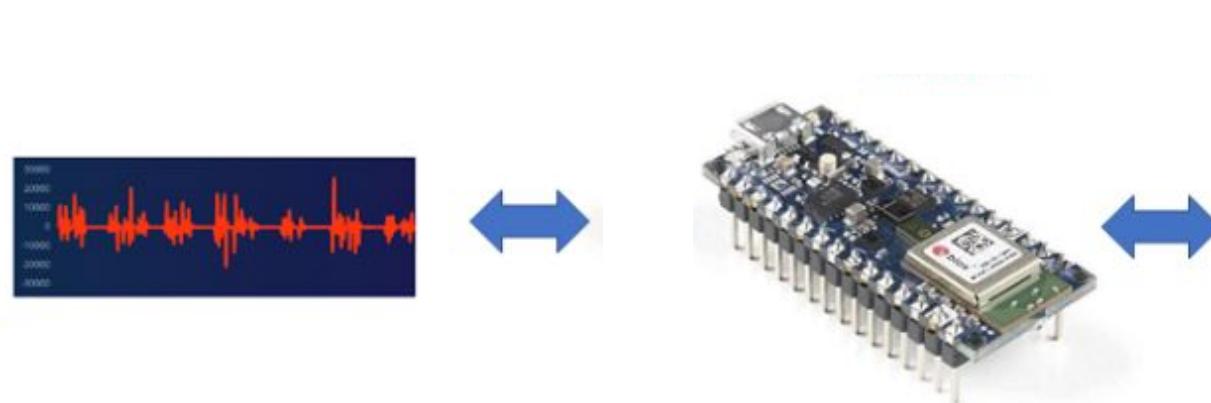
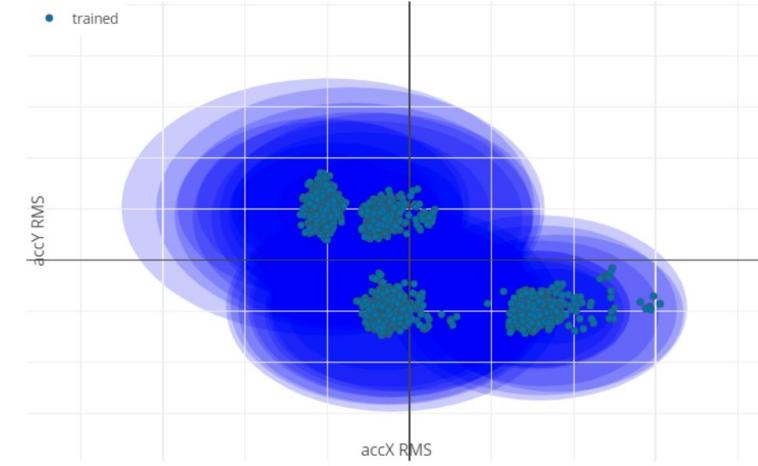
Human & Animal Sensing



Motion, radar, audio, PPG, ECG

- Health
- Consumer
- Industrial

Industrial – Anomaly Detection



IESTI01 2021.2 - Final Group Project: Bearing Failure Detection

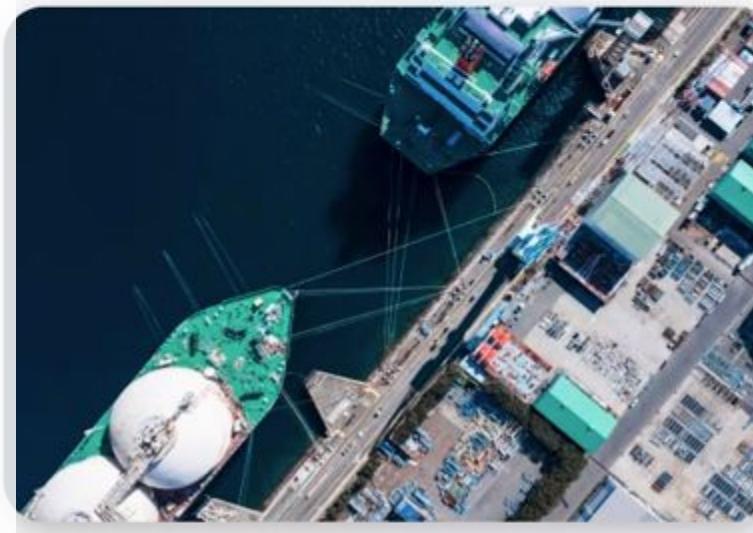
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- Infrastructure
- Buildings
- **Agriculture**

 **EDGE IMPULSE**

Human & Animal Sensing



Motion, radar, audio, PPG, ECG

- Health
- Consumer
- Industrial

Agriculture - Cow Monitoring

Using the Internet of Things for Agricultural Monitoring

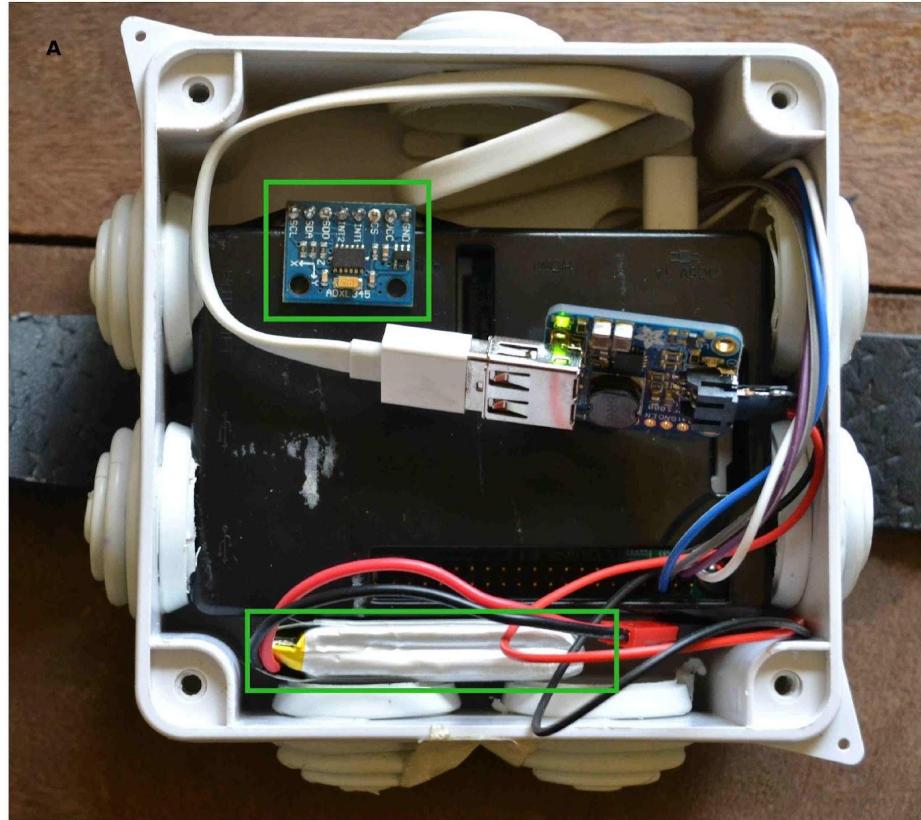
"We aim to deploy a variety of sensors for agricultural monitoring. One of the projects involves using **accelerometer sensors** to monitor activity levels in dairy cows with a view to determining when the cows are on heat or when they are sick."



Ciira wa Maina, Ph.D.

Senior Lecturer
Department of Electrical and Electronic Engineering
Dedan Kimathi University of Technology
Nyeri Kenya
Email: ciira.maina@dkut.ac.ke

Kenia



<https://sites.google.com/site/cwamainadekut/research>



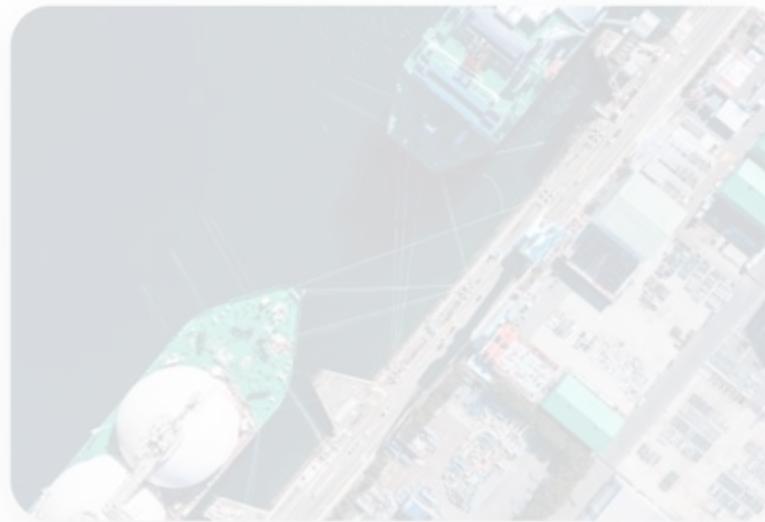
Predictive Maintenance



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Asset Tracking & Monitoring



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- Agriculture

Human & Animal Sensing



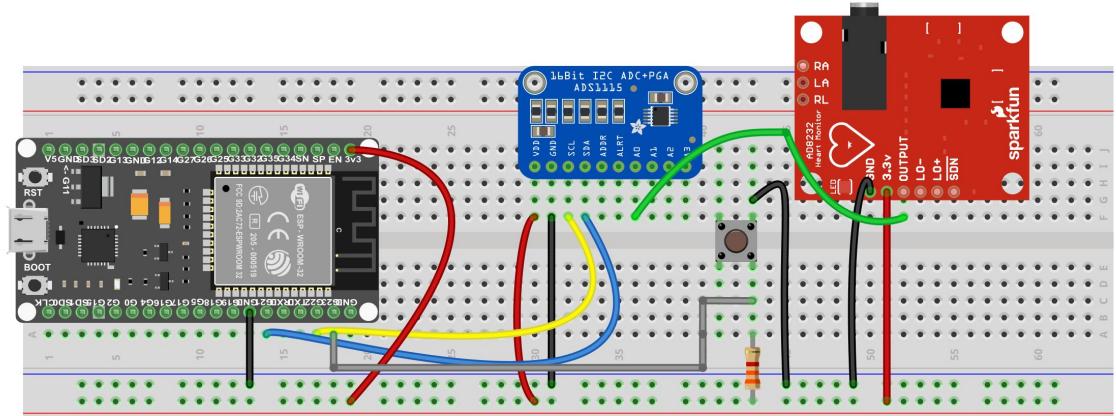
Motion, radar, audio, PPG, ECG

- **Health**
- Consumer
- Industrial

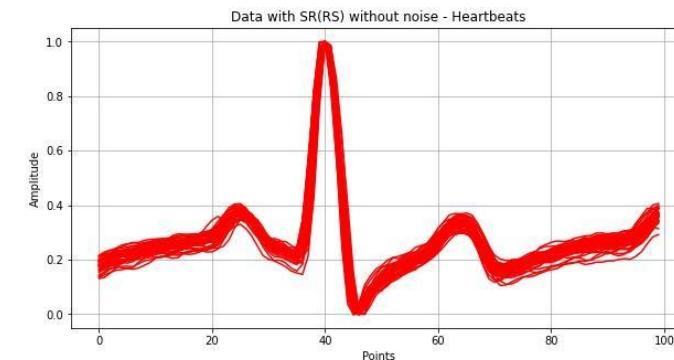
Health - Human Sensing



[Atrial Fibrillation Detection on ECG using TinyML](#)
Silva et al. UNIFEI 2021



fritzing



Guilherme Silva
Engenheiro - UNIFEI

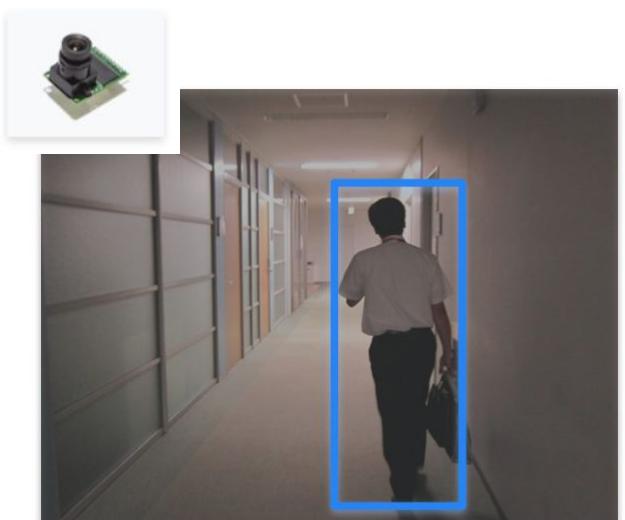
Sound



Vibration



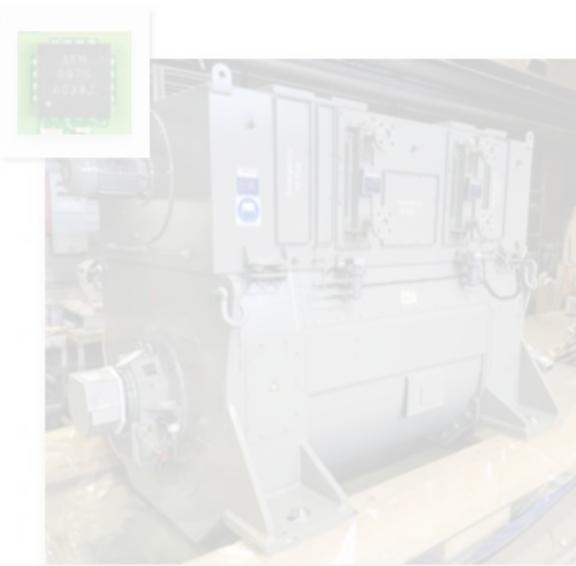
Vision



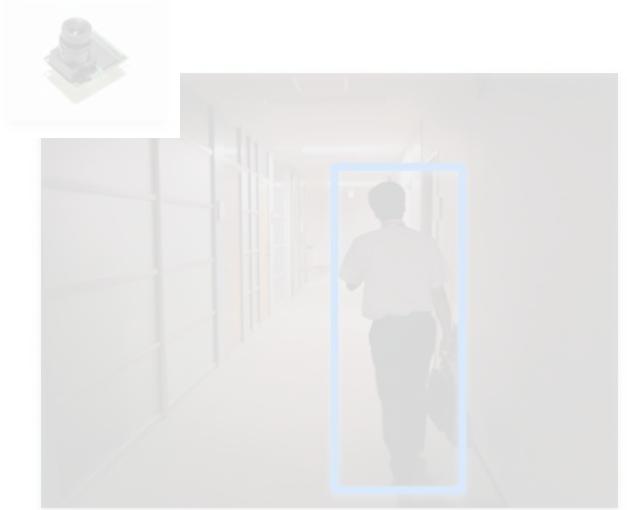
Sound



Vibration



Vision





Moez Altayeb
University of Khartoum, Sudan
ICTP, Trieste, Italy
mohedahmed@hotmail.com

ABSTRACT

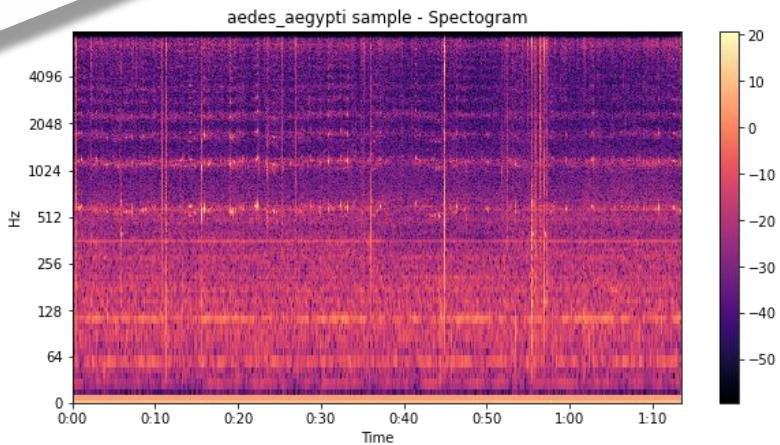
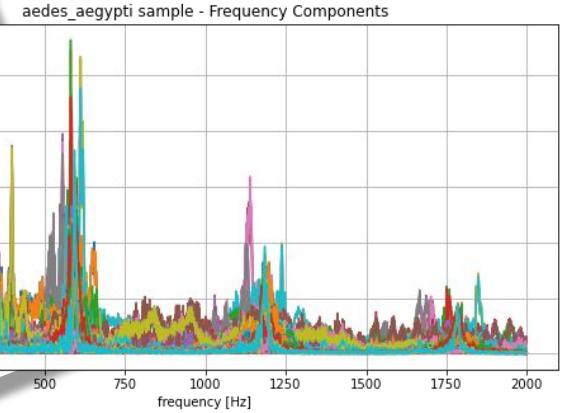
Every year more than one billion people are infected and more than one million people die from vector-borne diseases including malaria, dengue, zika and chikungunya. Mosquitoes are the best known disease vector and are geographically spread worldwide. It is important to raise awareness of mosquito proliferation by monitoring their incidence, especially in poor regions. Acoustic detection of mosquitoes has been studied for long and ML can be used to automatically identify mosquito species by their wingbeat. We present a prototype solution based on an openly available dataset on the Edge Impulse platform and on three commercially-available TinyML devices. The proposed solution is low-power, low-cost and can run without human intervention in resource-constrained areas. This insect monitoring system can reach a global scale.

Classifying mosquito wingbeat sound using TinyML

Marcelo Rovai
Universidade Federal de Itajubá
Itajubá, Brazil
rovai@unifei.edu.br

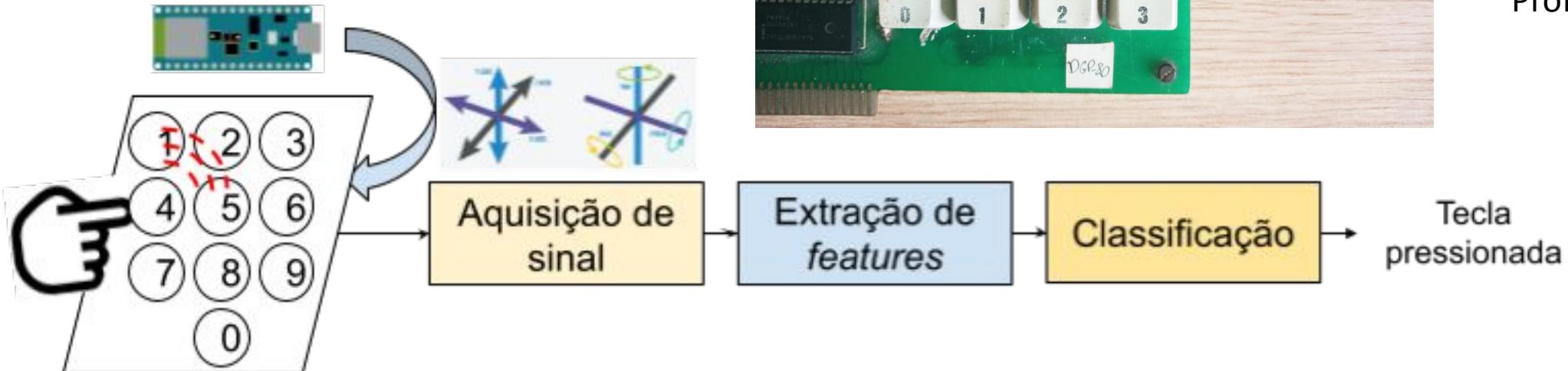
Marco Zennaro
ICTP
Trieste, Italy
mzennaro@ictp.it

affected. People from poor communities with little access to health care and clean water sources are also at risk. Although anti-malarial drugs exist, there's currently no malaria vaccine. Vector-borne diseases also exacerbate poverty. Illness prevent people from working and supporting themselves and their families, impeding economic development. Countries with intensive malaria have much lower income levels than those that don't have malaria. Countries affected by malaria turn to control rather than elimination. Vector control means decreasing contact between humans and disease carriers on an area-by-area basis. It is therefore of great interest to be able to detect the presence of mosquitoes in a specific area. This paper presents an approach based on TinyML and on embedded devices.

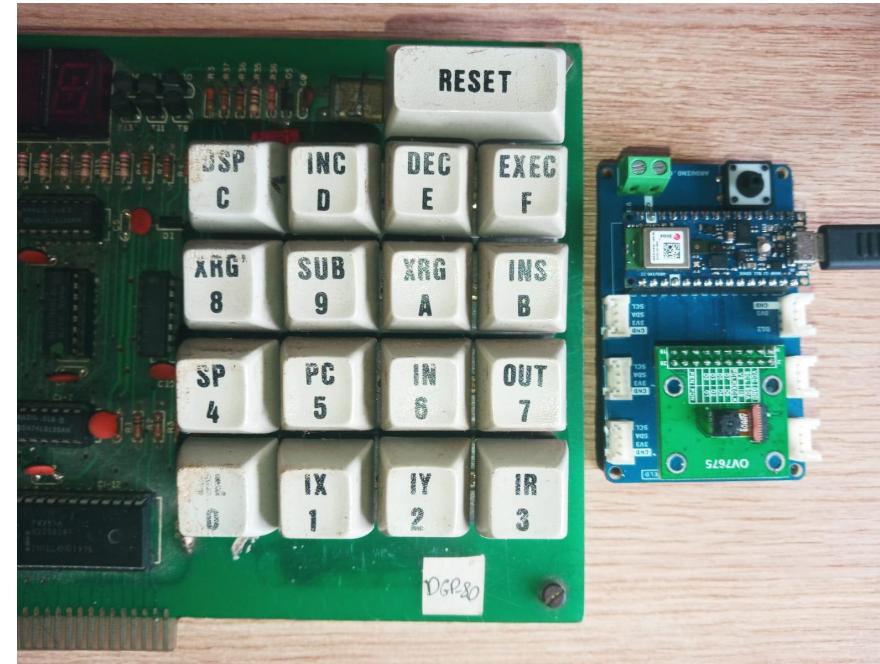


<https://github.com/Mjrovai/wingbeat-mosquito-tinyml>

Key Stroke Detection



IESTI01 2021.2 – Final Group Project Key Stroke Detection



Renan Castro
Professor IFESP

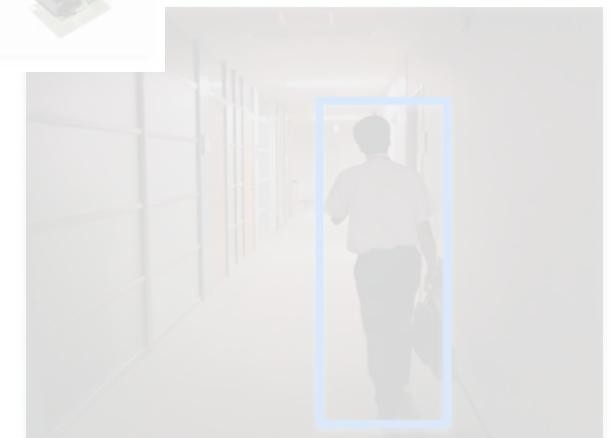
Sound



Vibration



Vision



Predict and classify common Elephant behavior



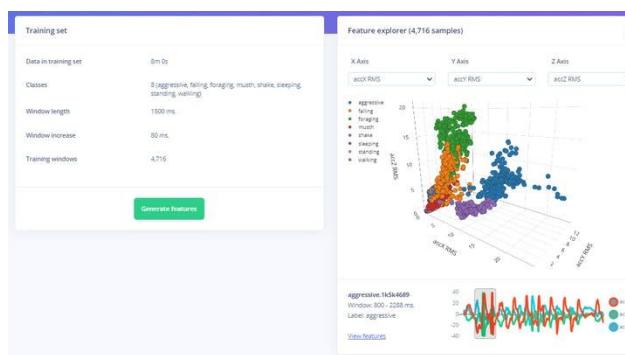
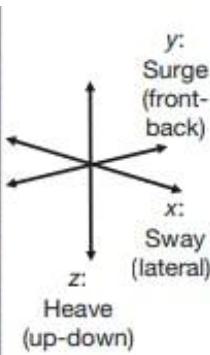
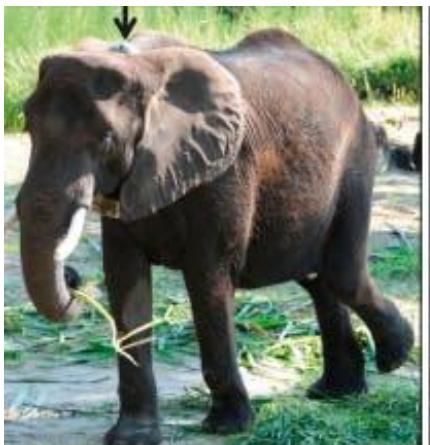
Aggressive



Standing



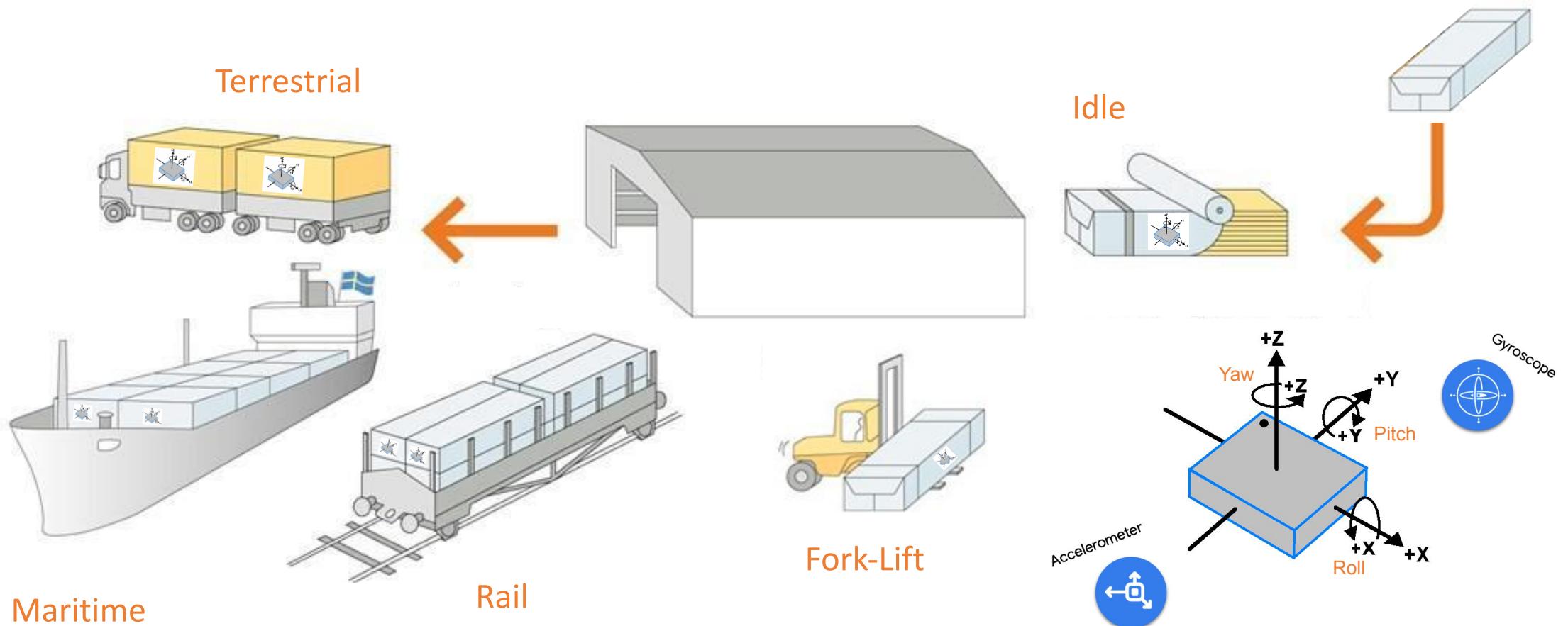
Sleeping



https://www.hackster.io/dhruvsheth_electet-tinyml-and-iot-based-smart-wildlife-tracker-c03e5a



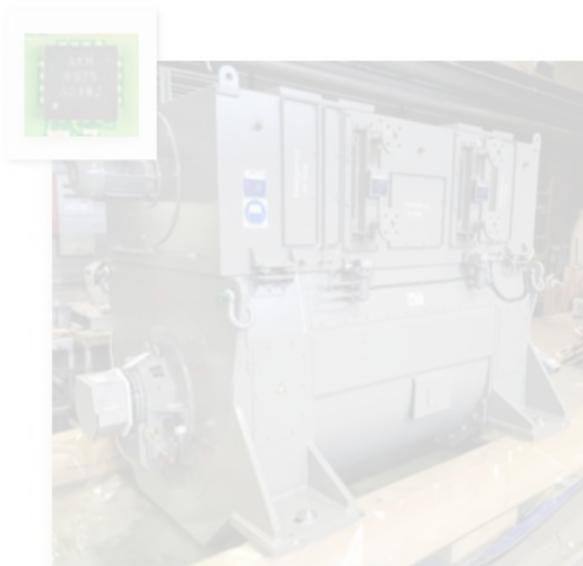
Mechanical Stresses in Transport



Sound



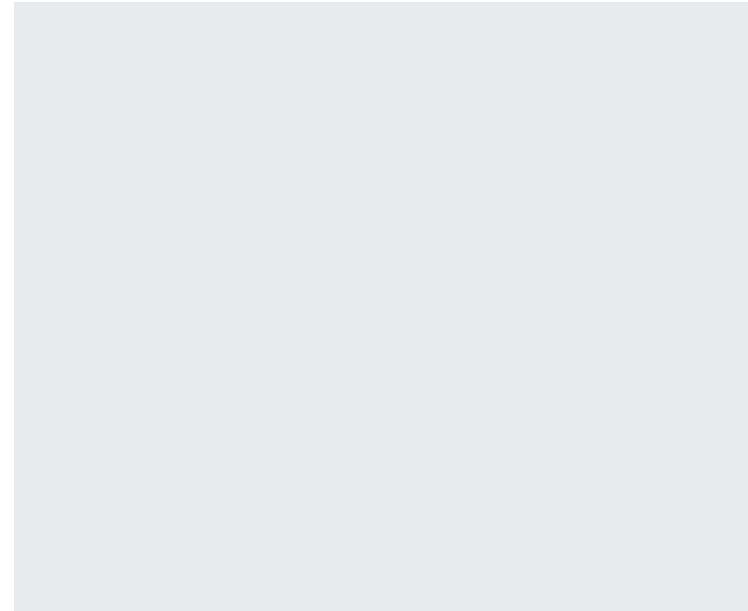
Vibration



Vision



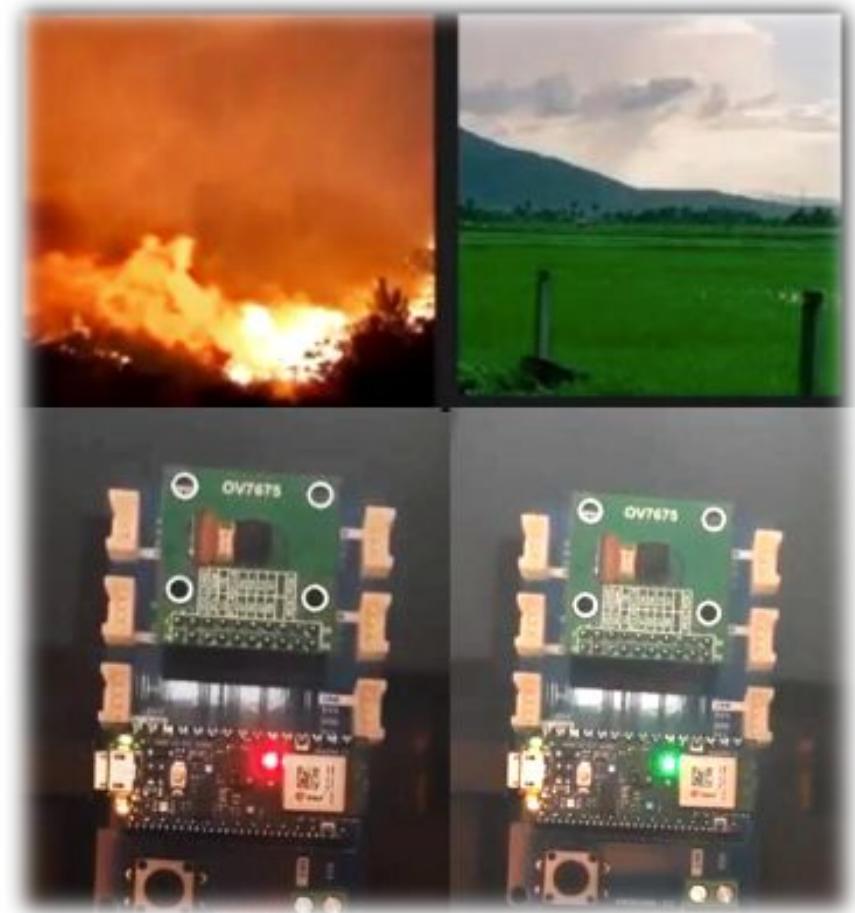
Person Detection



Forest Fire Detection



[TinyML Aerial Forest Fire Detection](#)



[IESTI01 - Forest Fire Detection – Proof of Concept](#)

Coffee Disease Classification



<https://www.hackster.io/Yukio/coffee-disease-classification-with-ml-b0a3fc>

Introdução

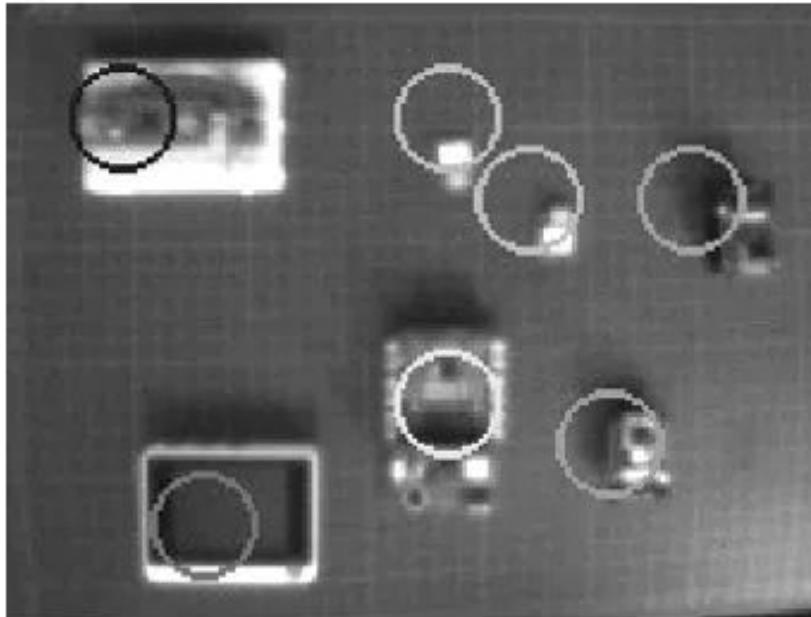
O Brasil é responsável por 50% do café exportado globalmente, o que é uma atividade importante para o país; geralmente a análise e classificação de doenças em plantas é feita manualmente, que não são acessíveis para pequenos produtores.

Com o aumento do poder de processamento das placas-mãe microcontroladas e processadores dedicados ao machine learning, a tarefa de embarcar todos morais tem-se tornado positiva em diversas áreas.



João Vitor Yukio Bordin Yamashita
Graduando em Engenharia Eletrônica pela UNIFEI

Detecting Objects using TinyML (FOMO)



```
***** espcam *****
x 70  y 150
x 130  y 170
*****
***** nano *****
x 70  y 110
*****
***** pico *****
x 150  y 30
*****
***** wio *****
x 50  y 50
*****
***** xiao *****
x 150  y 110
x 130  y 130
6.97512 fps
```

[EdgeAI made simple - Exploring Image Processing \(Object Detection\) on microcontrollers with Arduino Portenta, Edge Impulse FOMO, and OpenMV](#)

Other TinyML / MCUs Project Examples

Vision

- Image Classification with ESP32-CAM
- Image Classification with Portenta H7

[\[Doc\]](#)
[\[Doc\]](#)

Sound

- Listening Temperature with Nano 33

[\[Doc\]](#)

Vibration

- Motion Recognition with RPi Pico
- Gesture Recognition with Wio Terminal

[\[Doc\]](#)
[\[Doc\]](#)

TinyML Application

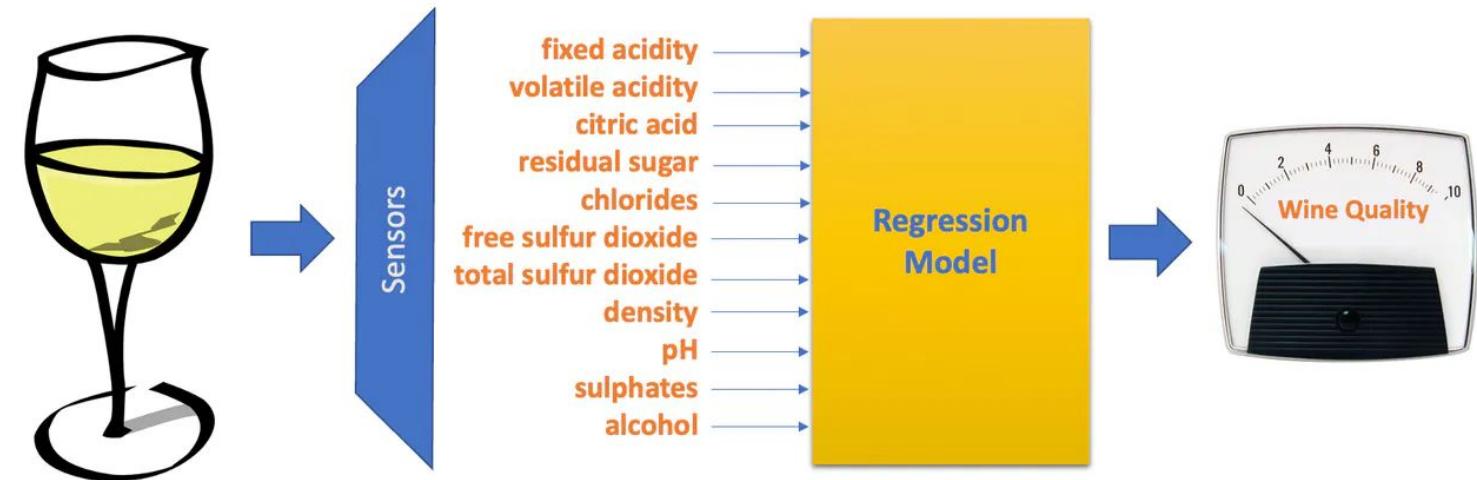
Examples (Regression)



Estimate Weight From a Photo
Using Visual Regression in
Edge Impulse



Regression on TinyML

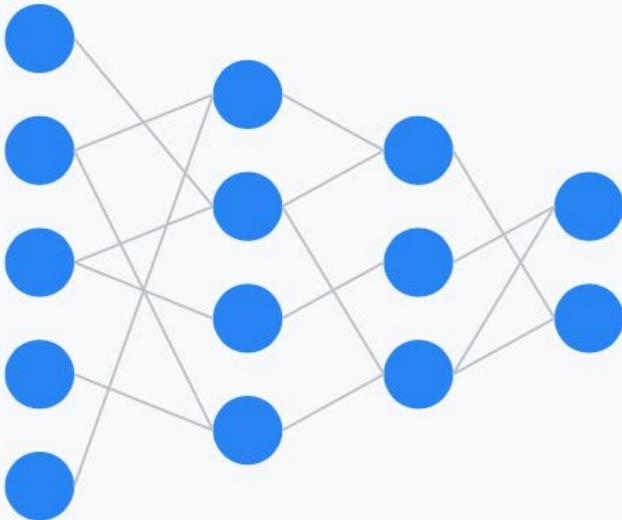


TinyML Made Easy: Exploring Regression - White Wine Quality

ML Lifecycle

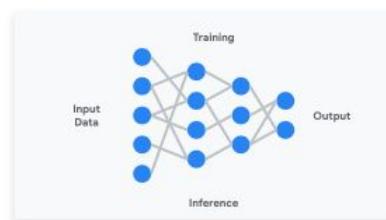
**Input
Data**

Training



Output

Inference



ML Code

Data Collection

Data Preprocessing

Debugging

Resource Management

Configuration

Data Verification

ML Code

Model Analysis

Serving Infrastructure

Automation

Feature Engineering

Process Management

Monitoring

Metadata Management

AI Infrastructure

Data Engineering

Model Engineering

Model Deployment

Product Analytics

Data Engineering

- Defining data **requirements**
- **Collecting** data
- **Labelling** the data
- Inspect and **clean** the data
- Prepare data for **training**
- **Augment** the data
- Add **more data**

AI Infrastructure

Data Engineering

Data Engineering

- Defining data **requirements**
- **Collecting** data
- **Labelling** the data
- Inspect and **clean** the data
- Prepare data for **training**
- **Augment** the data
- **Add more data**

AI Infrastructure

Data Engineering

Model Engineering

- **Training** ML models
- Improving training **speed**
- Setting **target** metrics
- **Evaluating** against metrics
- **Optimizing** model training
- Keeping up with **SOTA***

AI Infrastructure

Data Engineering

Model Engineering

* “**S**tate **o**f the **A**rt”

Model Deployment

- Model **conversion**
- **Performance** optimization
- **Energy-aware** optimizations
- **Security** and **privacy**
- **Inference** serving APIs
- **On-device** fine-tuning

AI Infrastructure

Data Engineering

Model Engineering

Model Deployment

Product Analysis

- Dashboards
- Field data evaluation
- Value-added for business
- Opportunities for advancement and improvements

AI Infrastructure

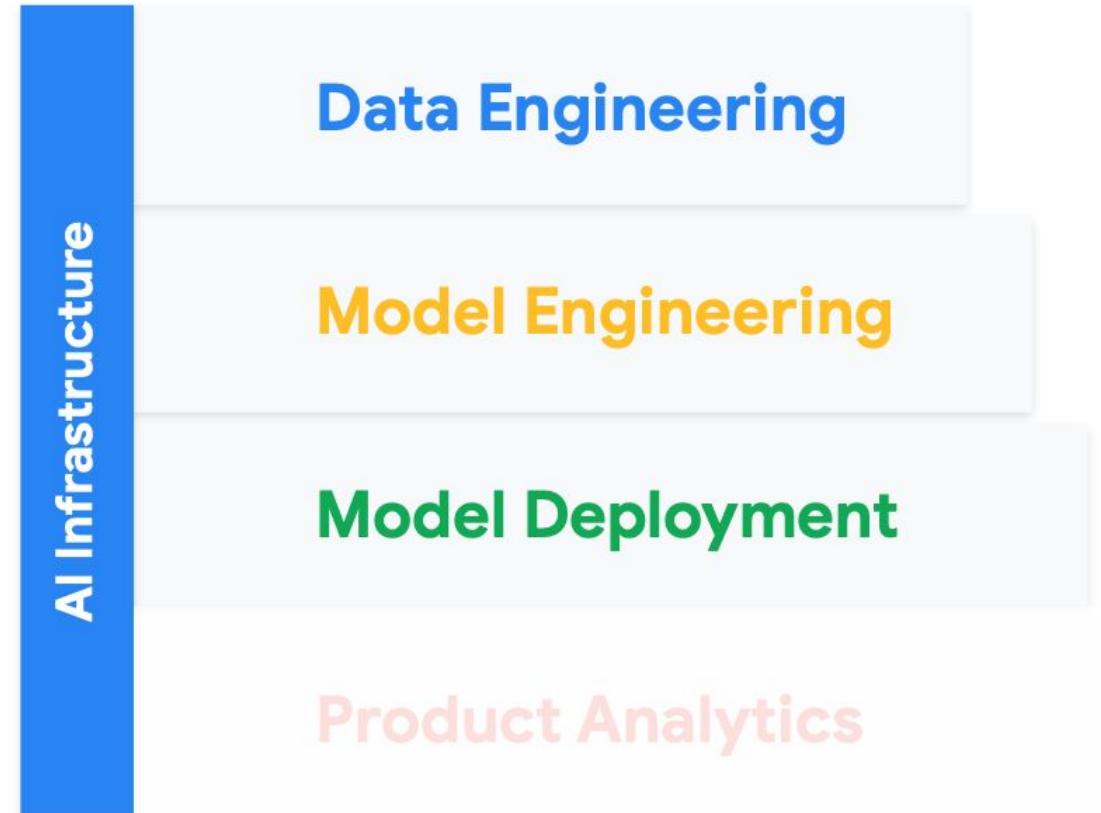
Data Engineering

Model Engineering

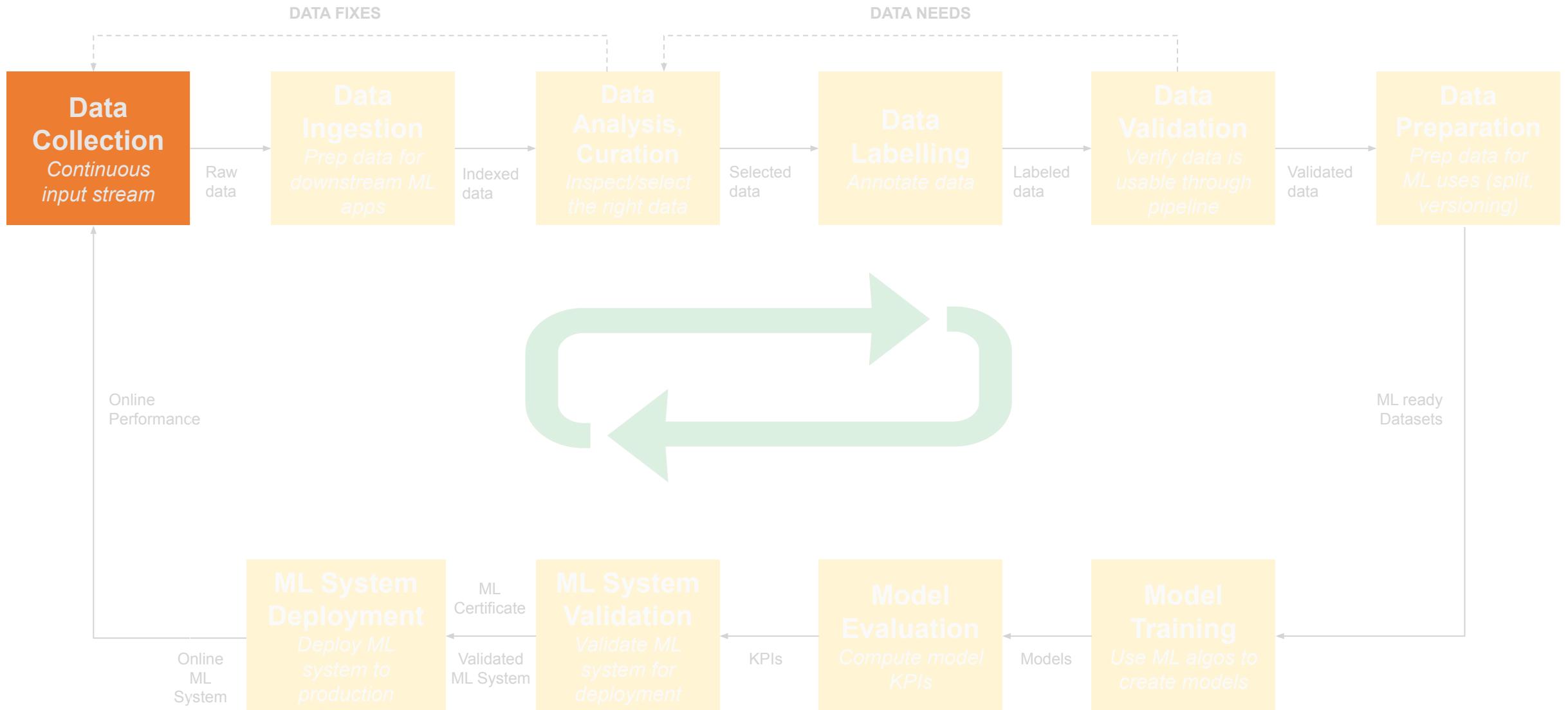
Model Deployment

Product Analytics

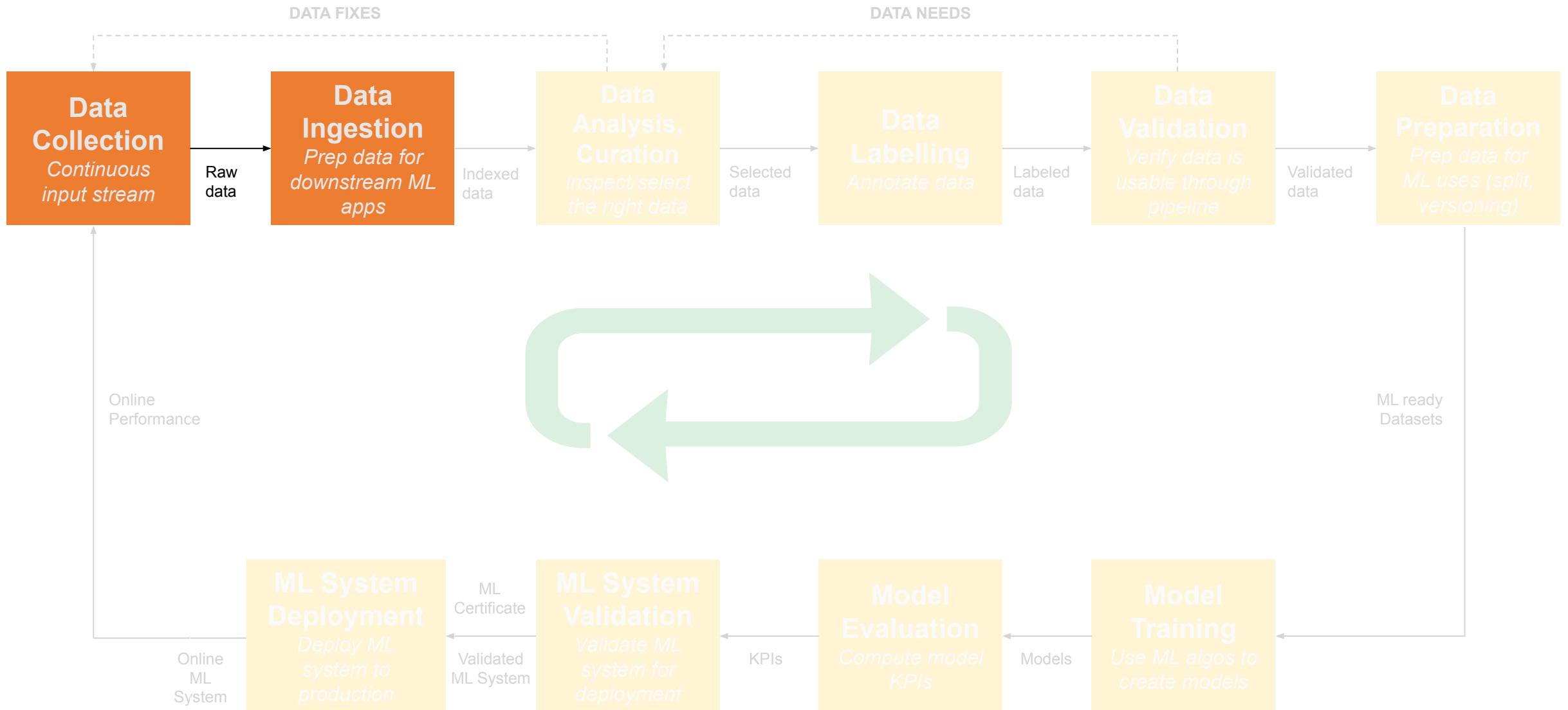
Focus in TinyML



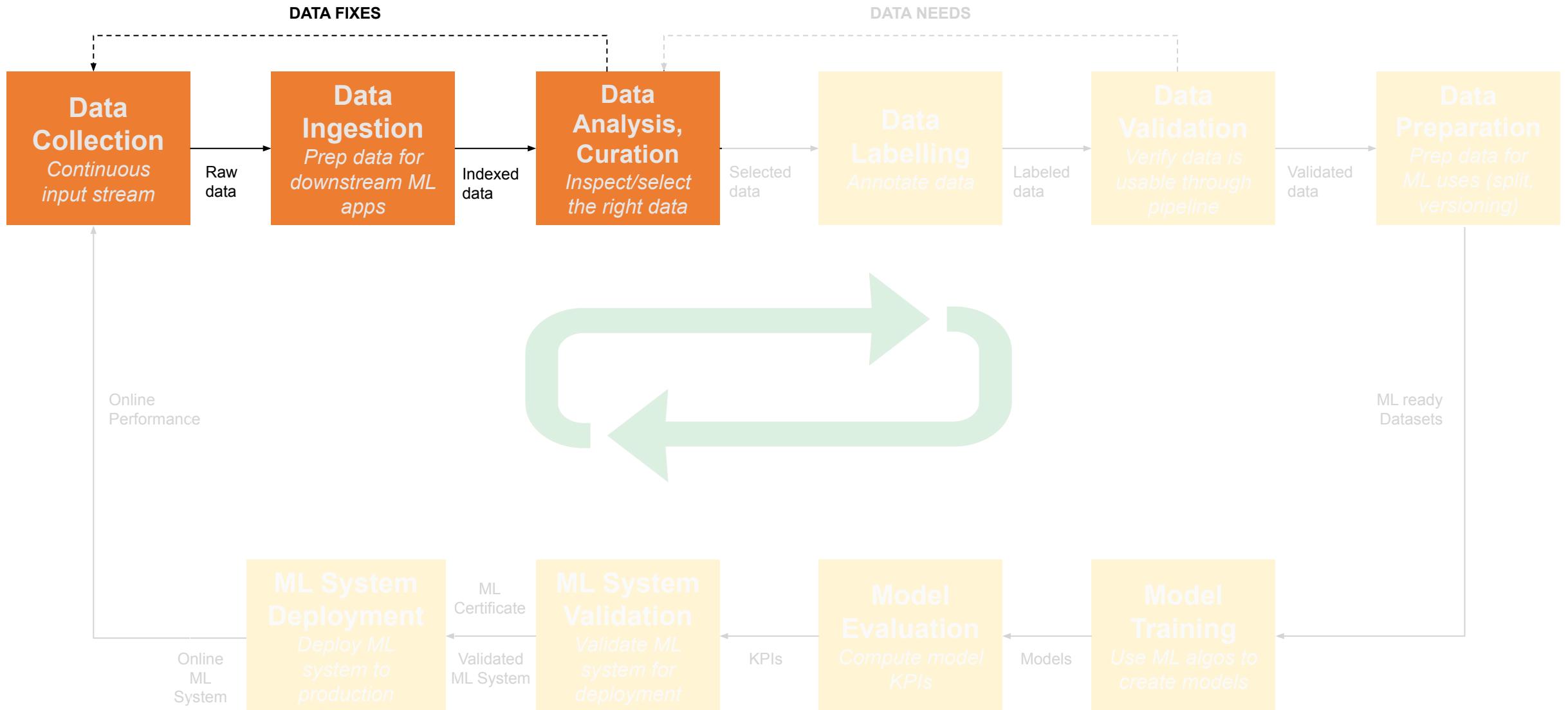
Life cycle of ML



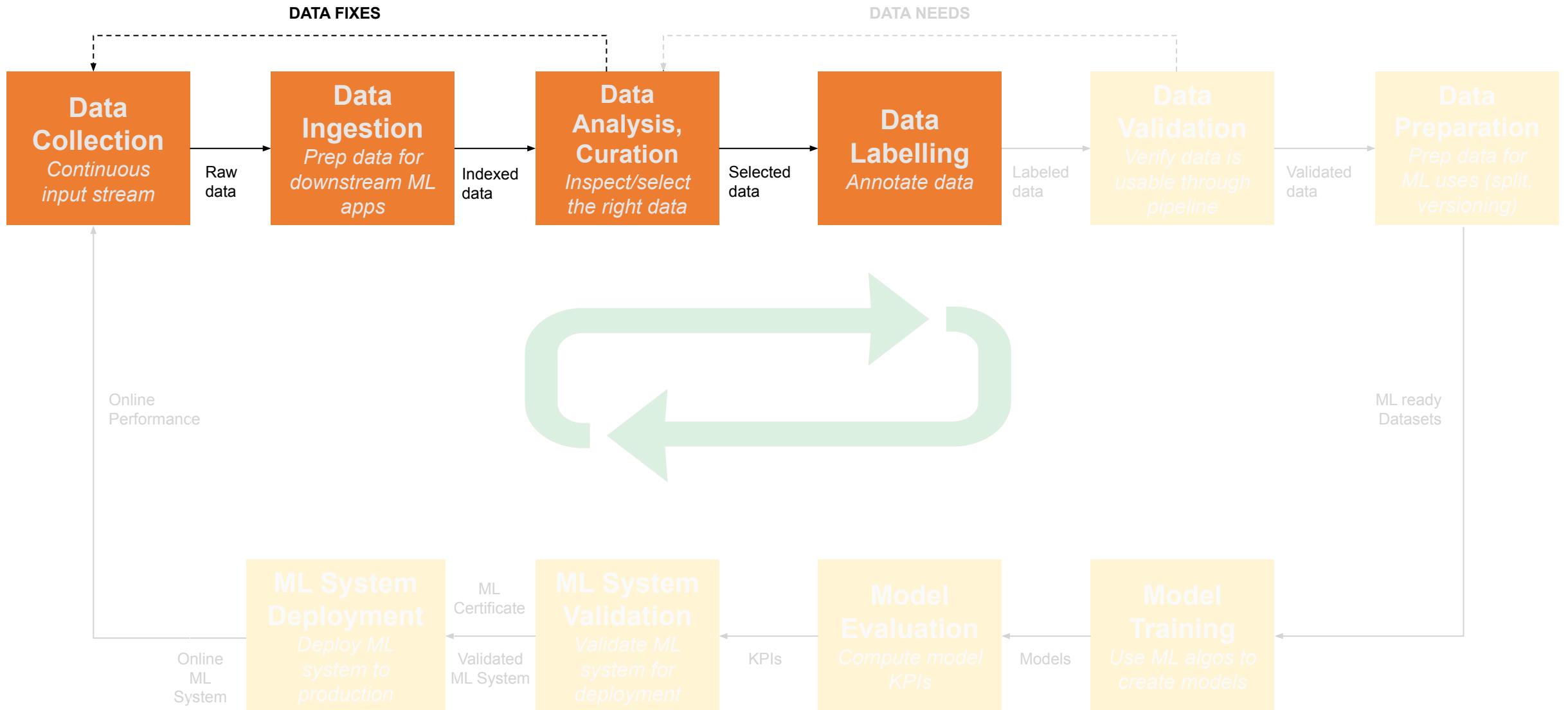
Life cycle of ML



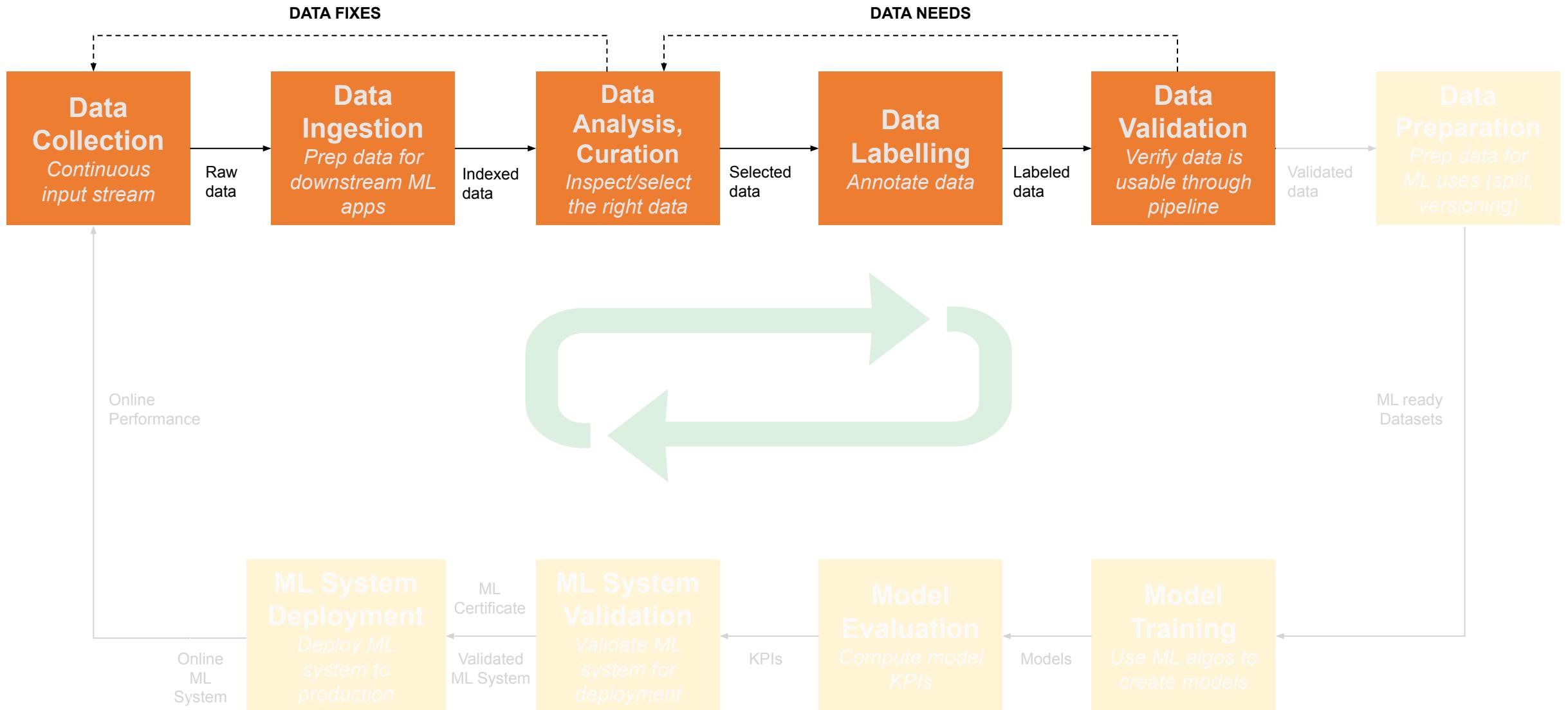
Life cycle of ML



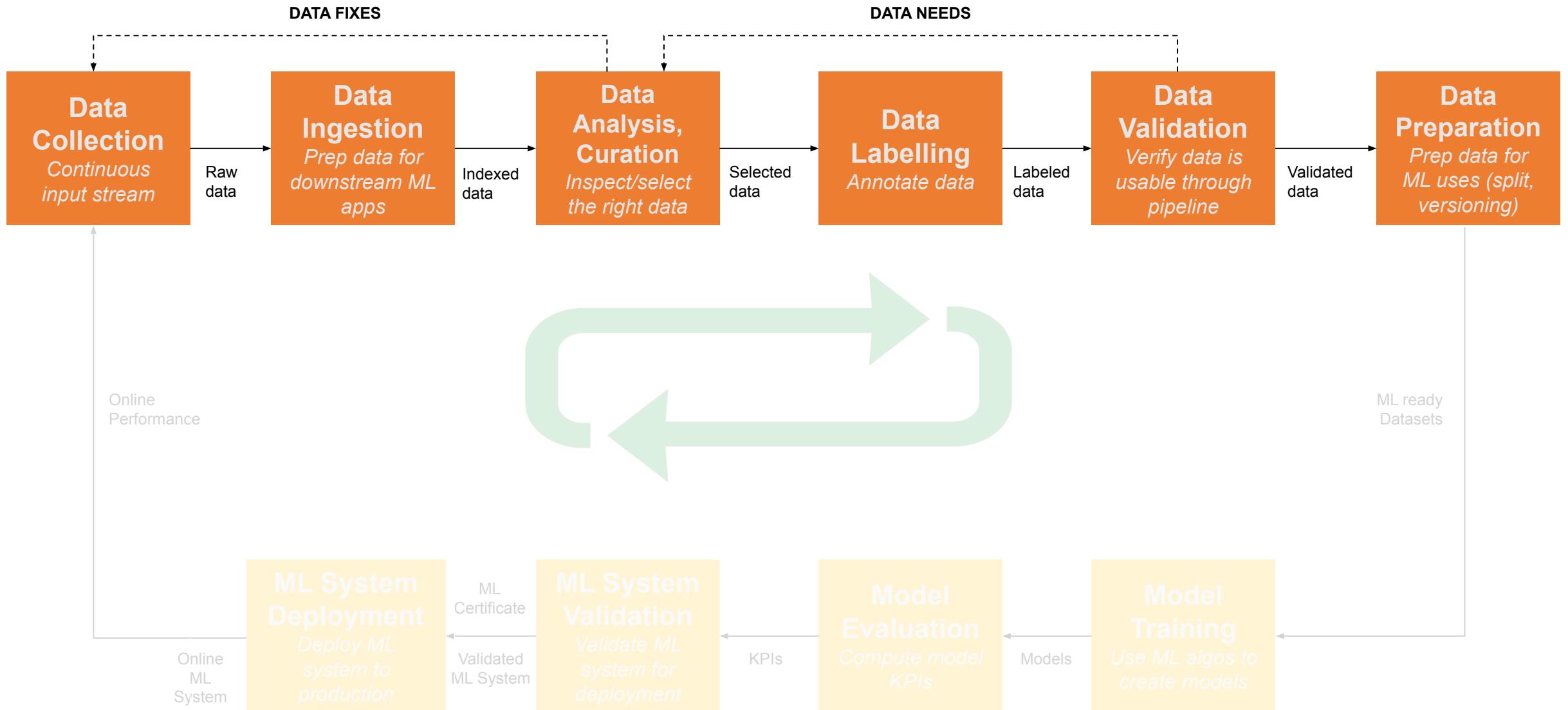
Life cycle of ML



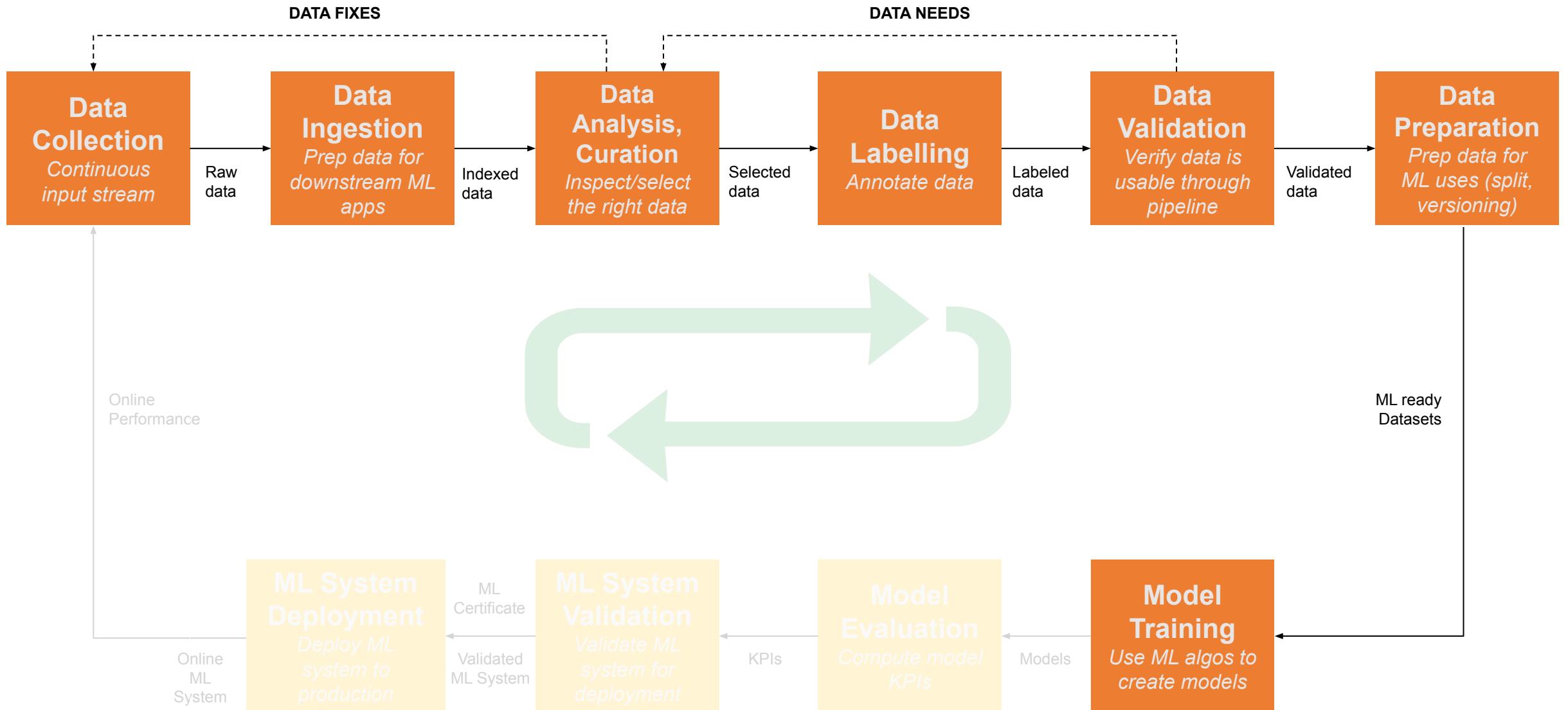
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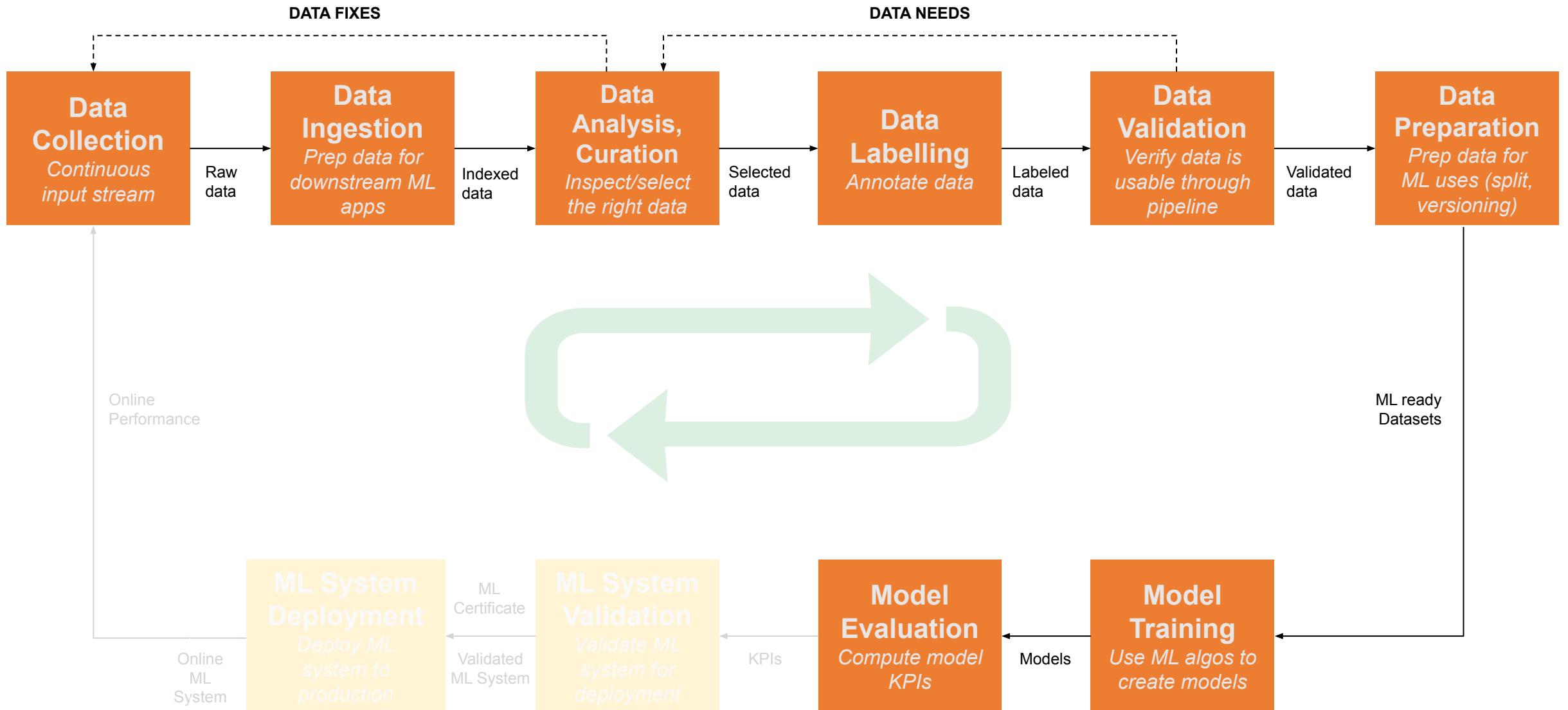
Life cycle of ML



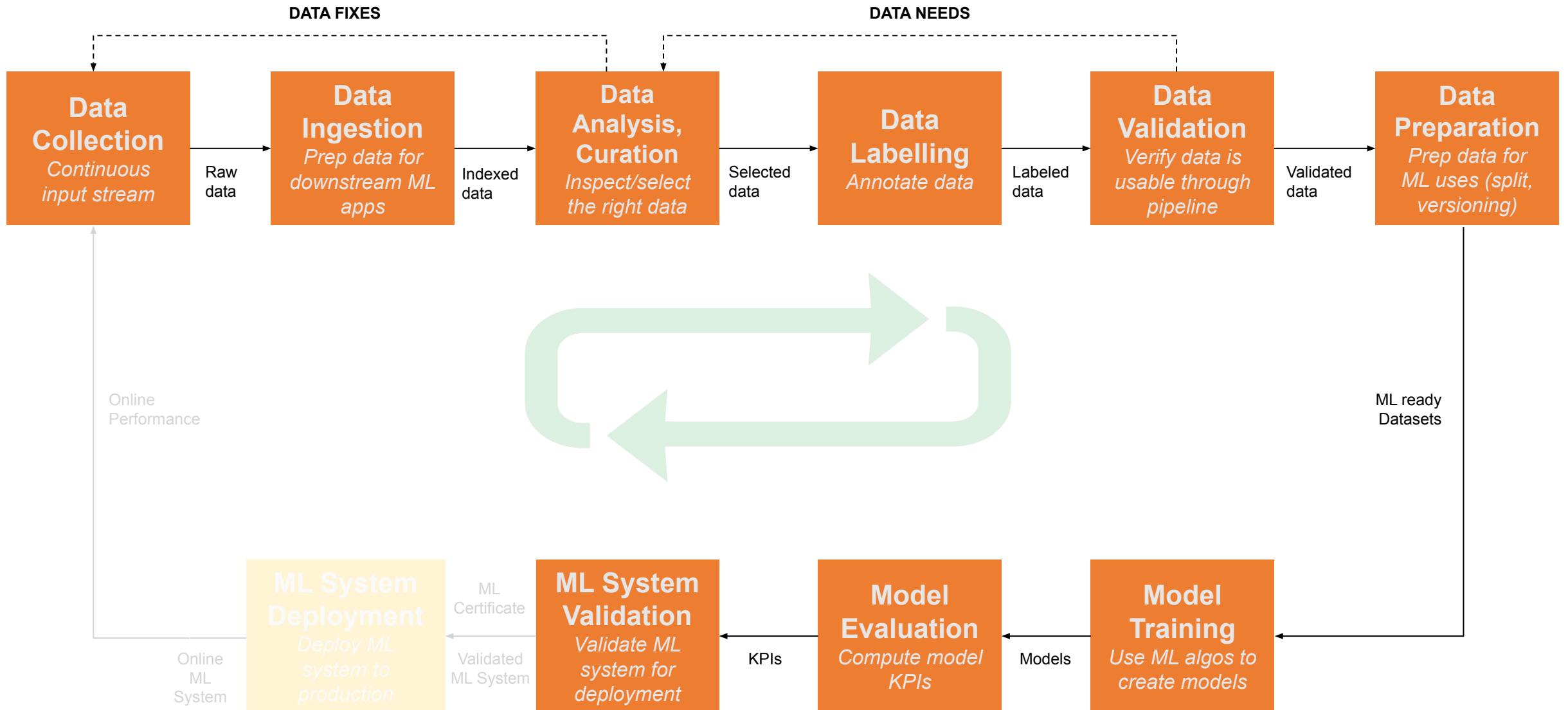
Life cycle of ML



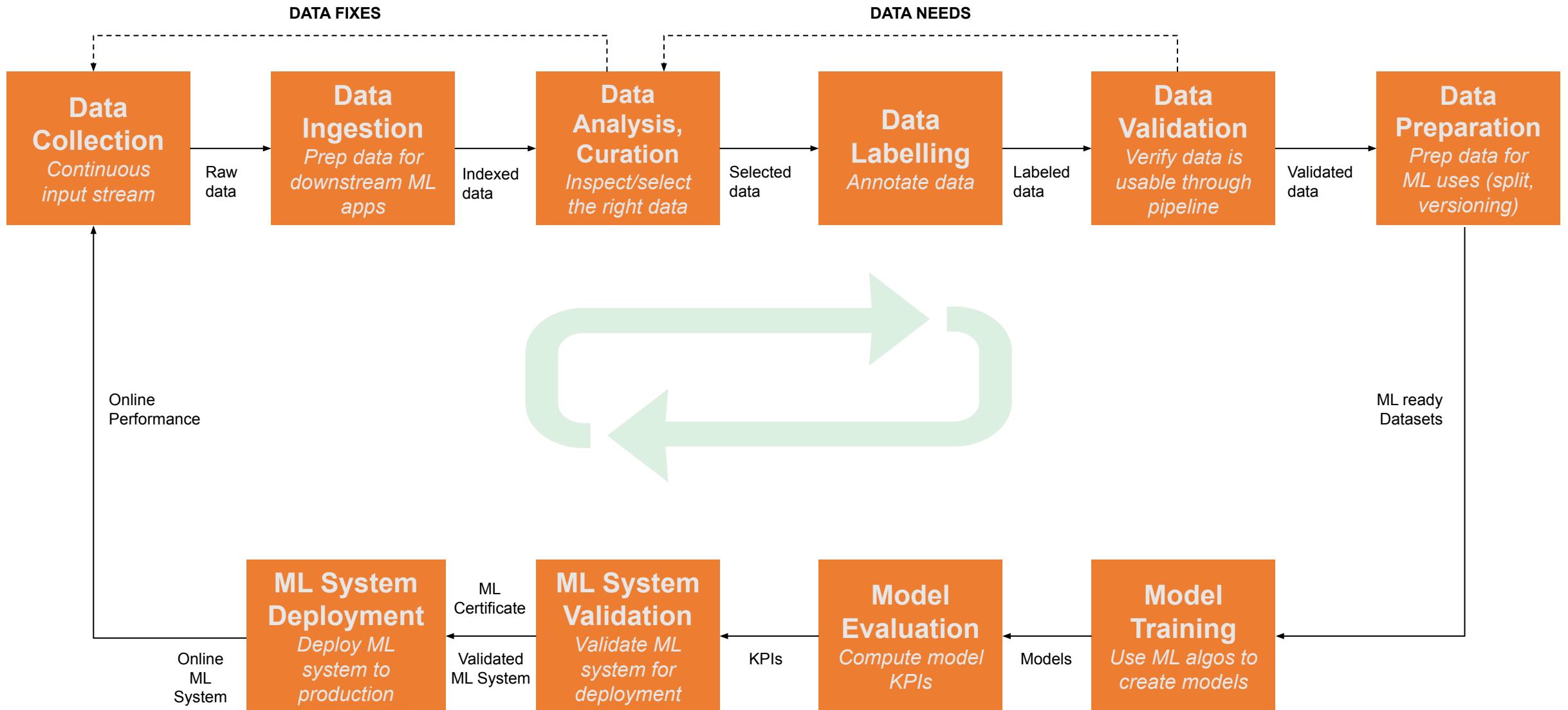
Life cycle of ML



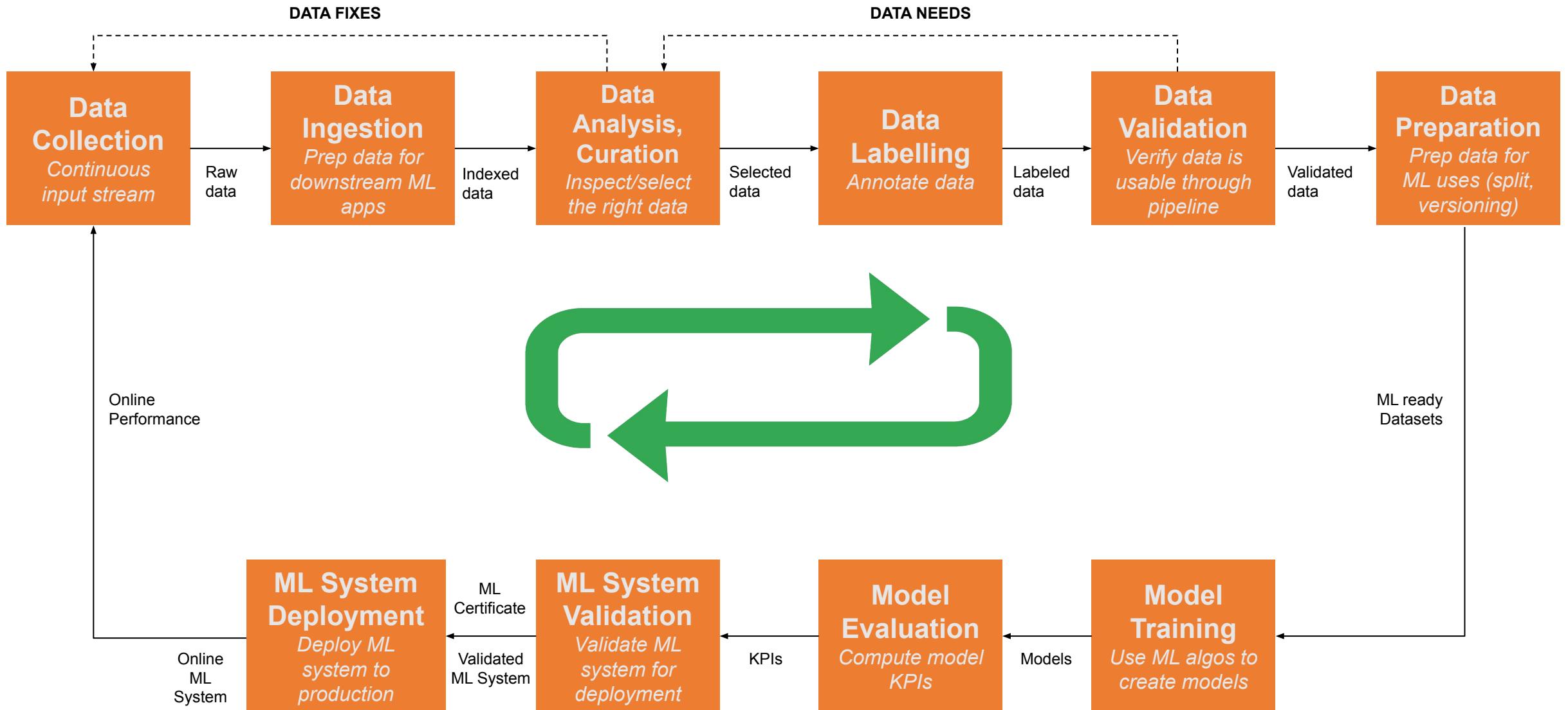
Life cycle of ML



Life cycle of ML



Life cycle of ML



ML Workflow

AI Infrastructure

Data Engineering

Model Engineering

Model Deployment

Product Analytics

Acoustic Sensors
Ultrasonic, Microphones,
Geophones, Vibrometers



Image Sensors
Thermal, Image



Motion Sensors
Gyroscope, Radar,
Accelerometer



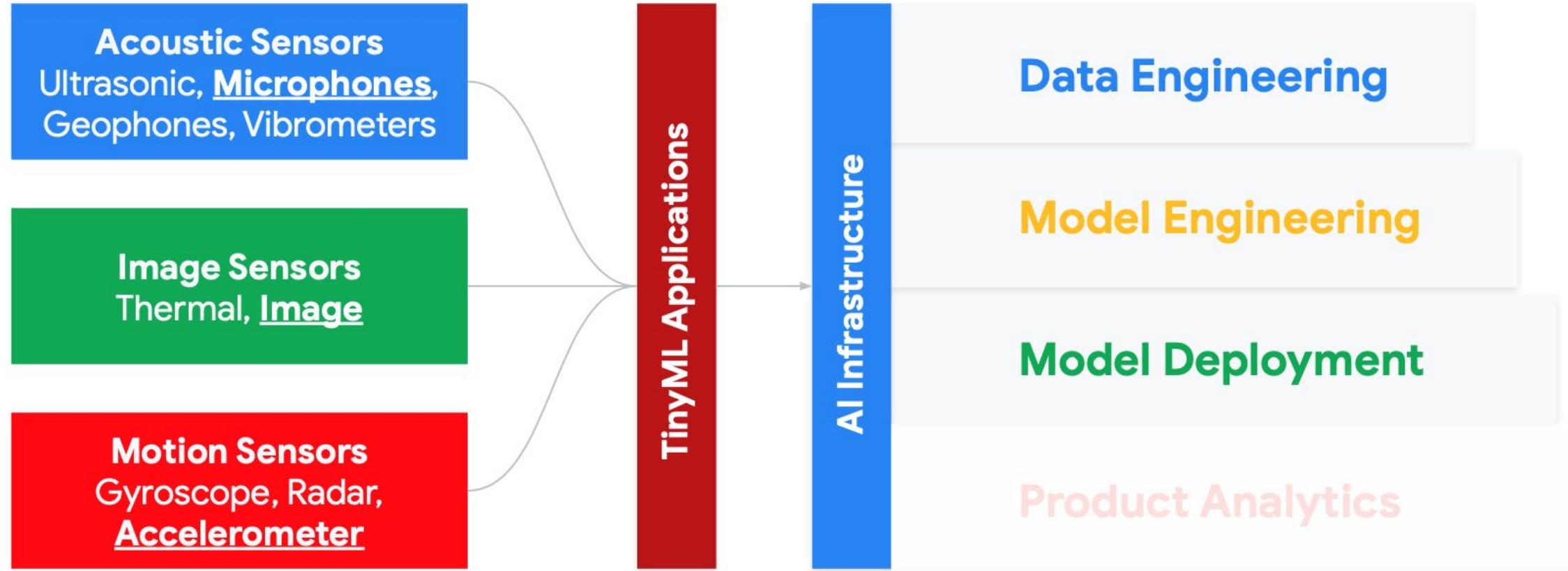
AI Infrastructure

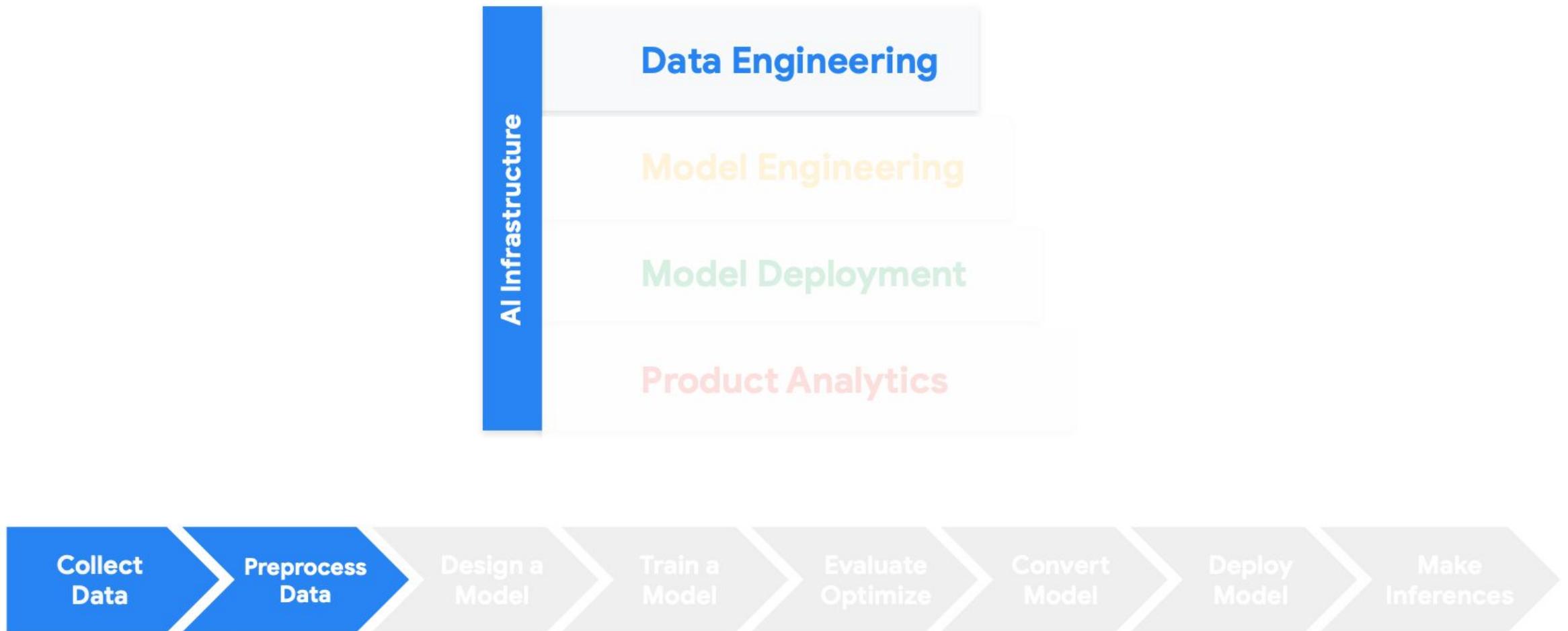
Data Engineering

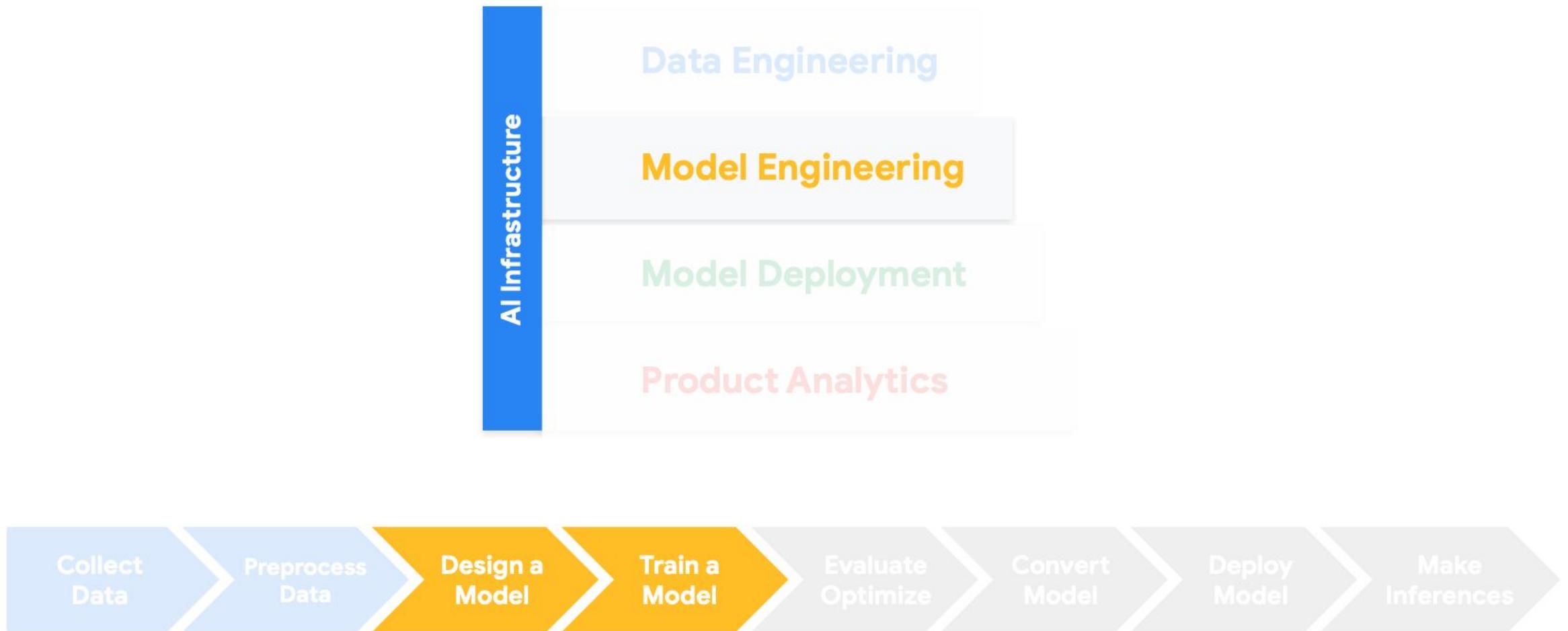
Model Engineering

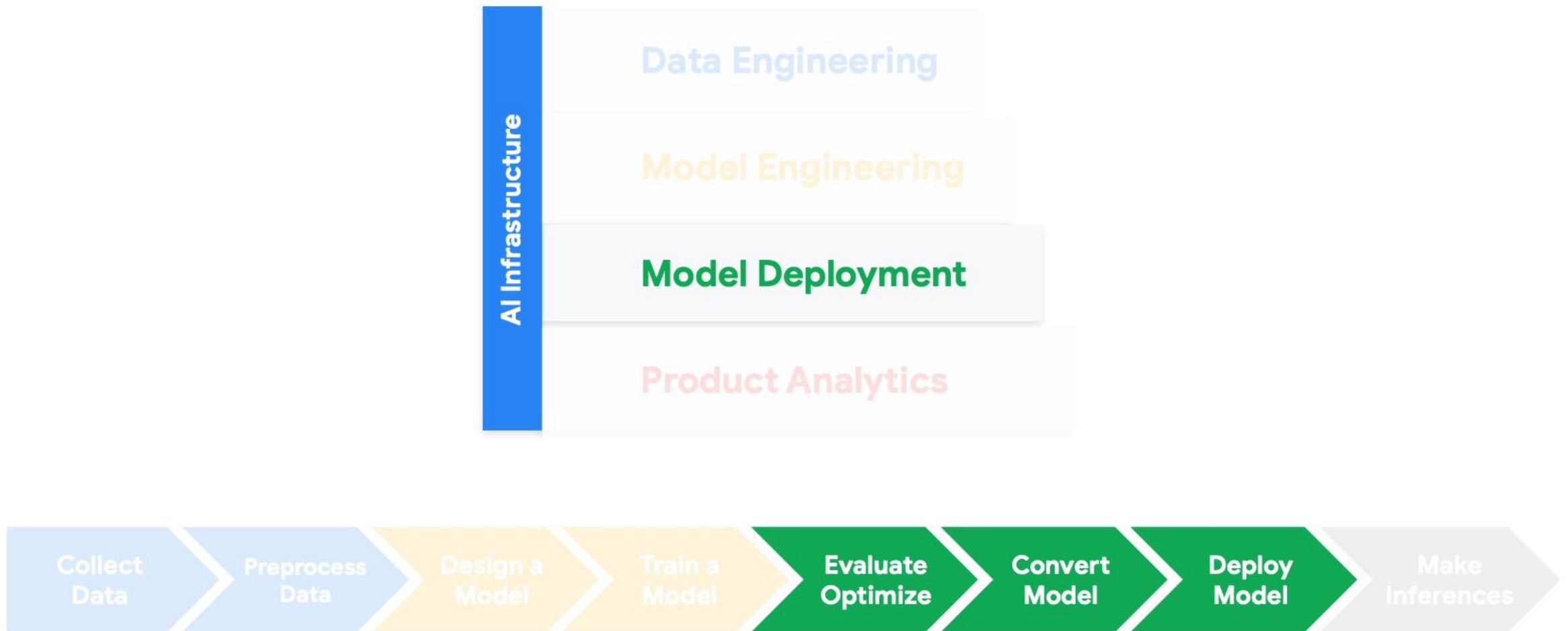
Model Deployment

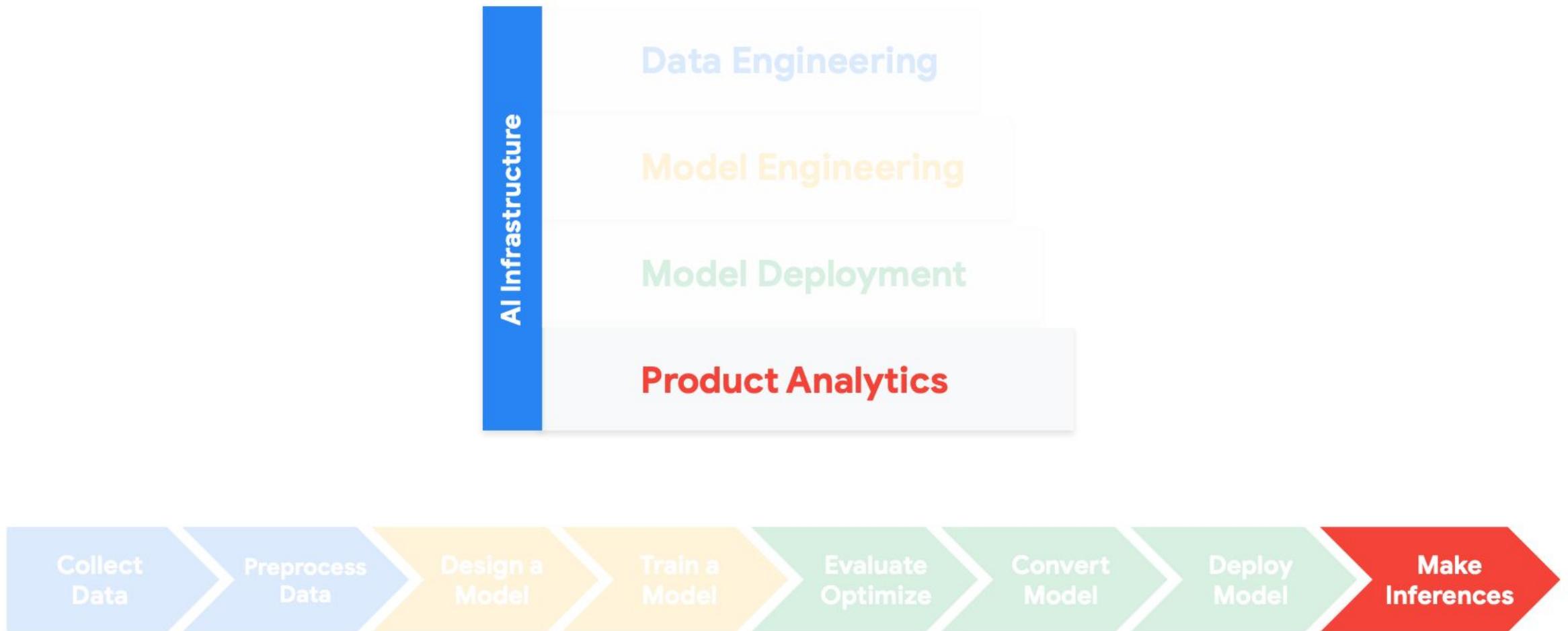
Product Analytics

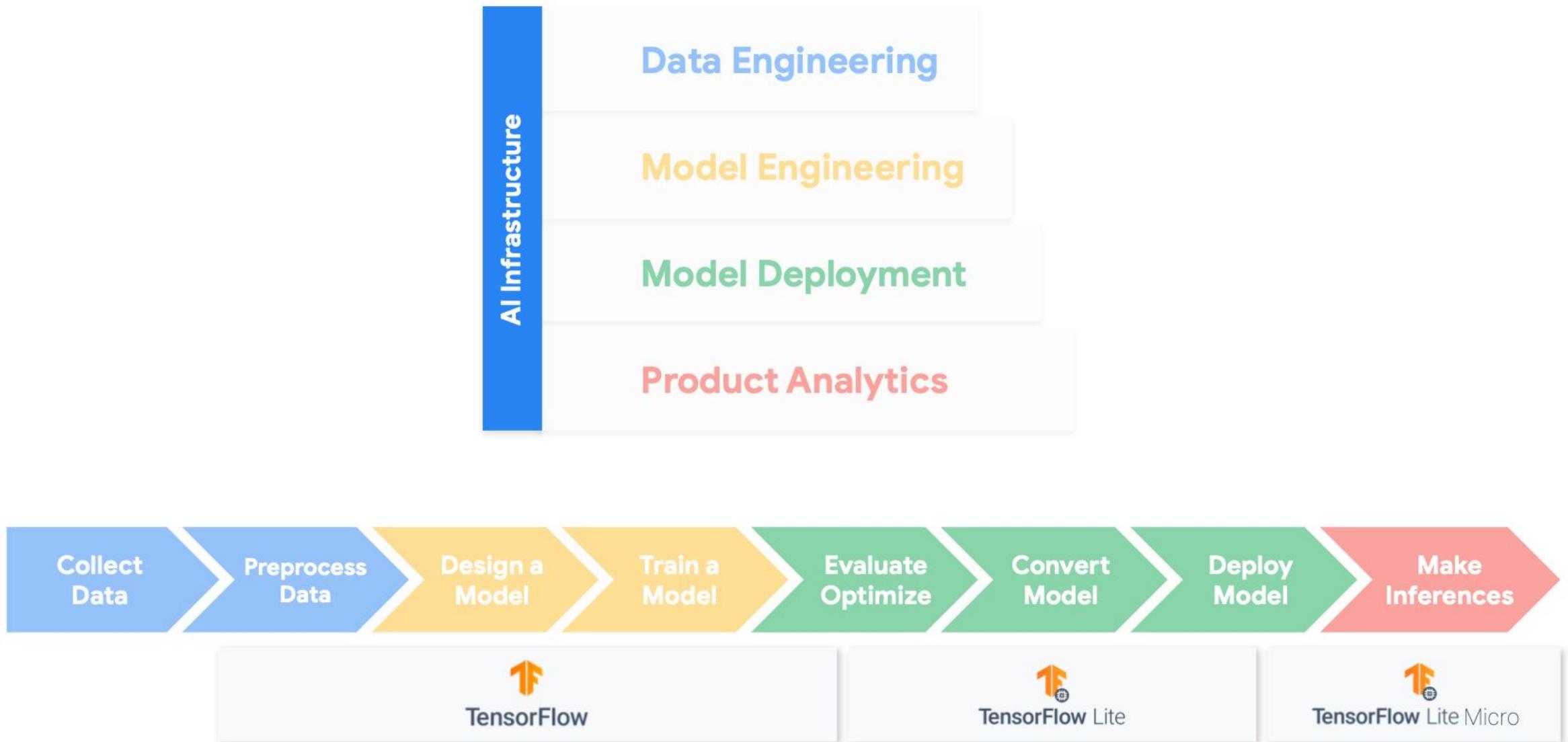


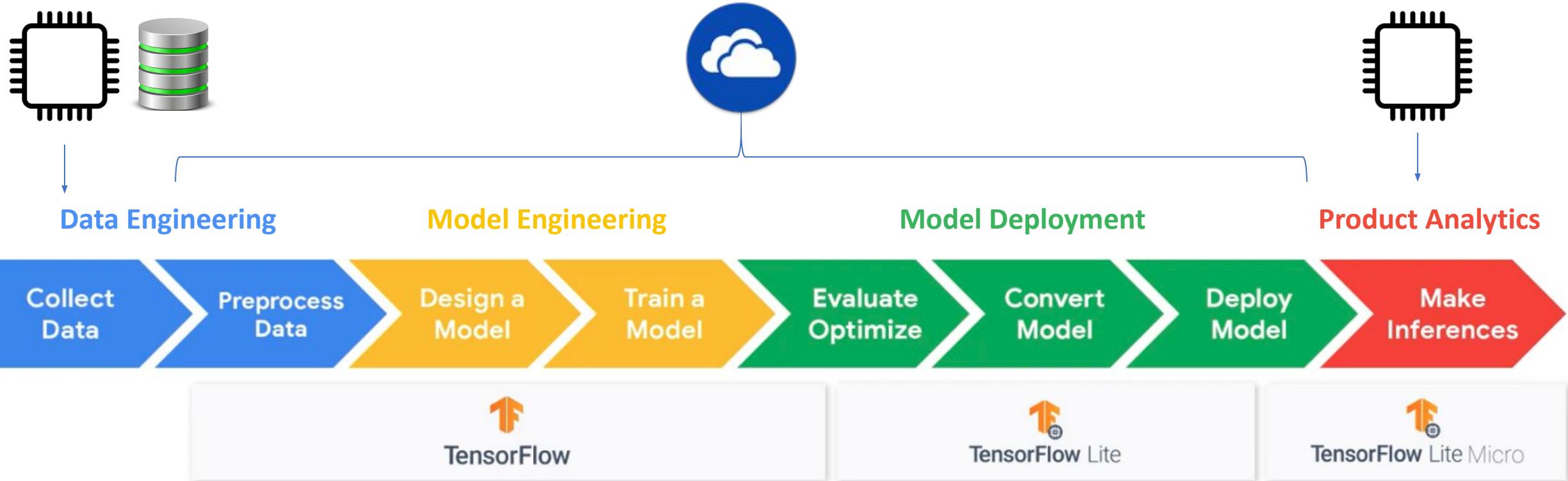


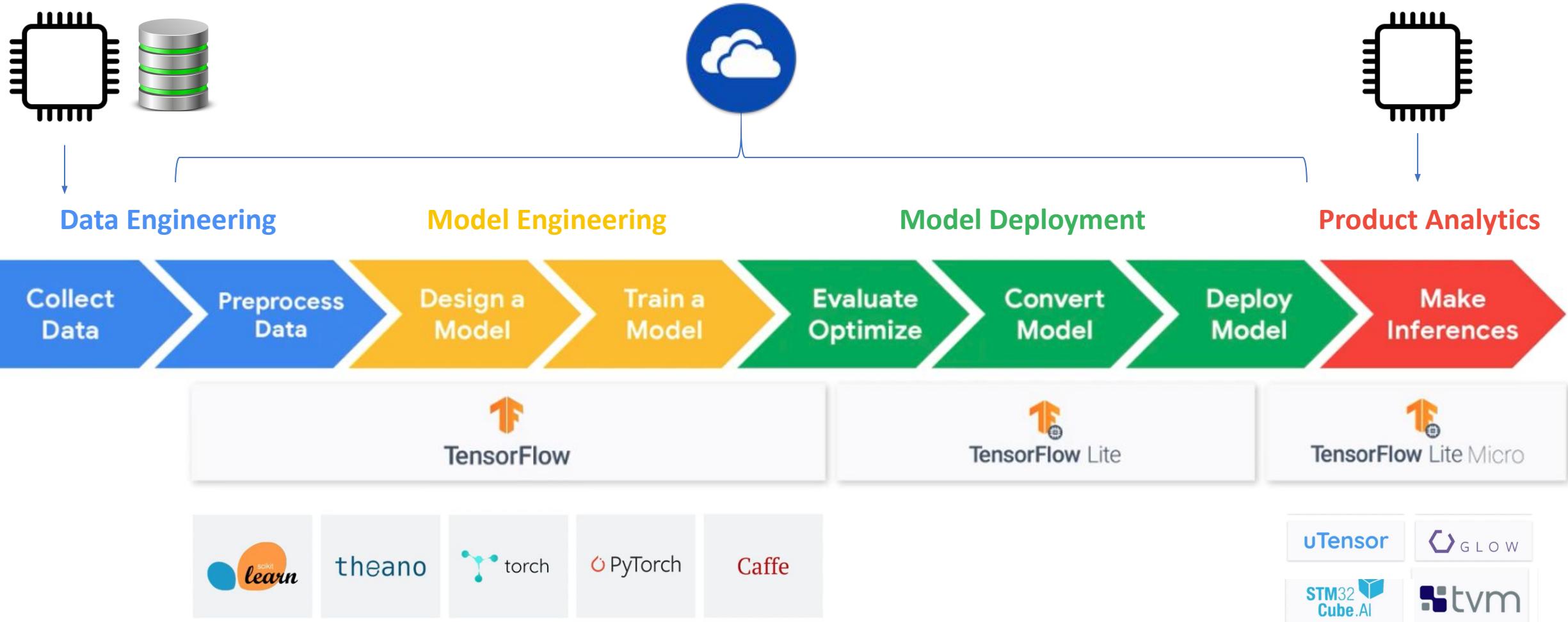


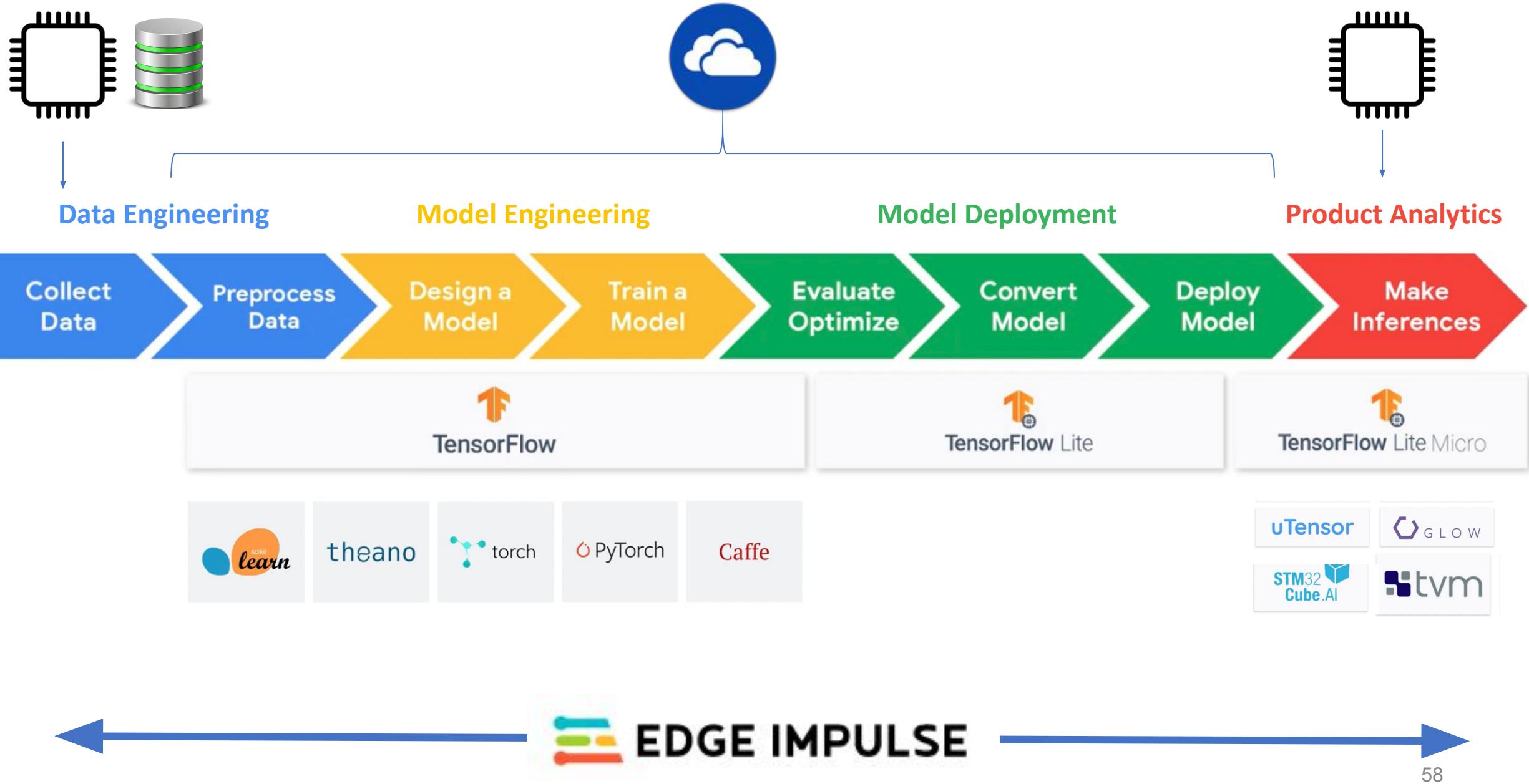


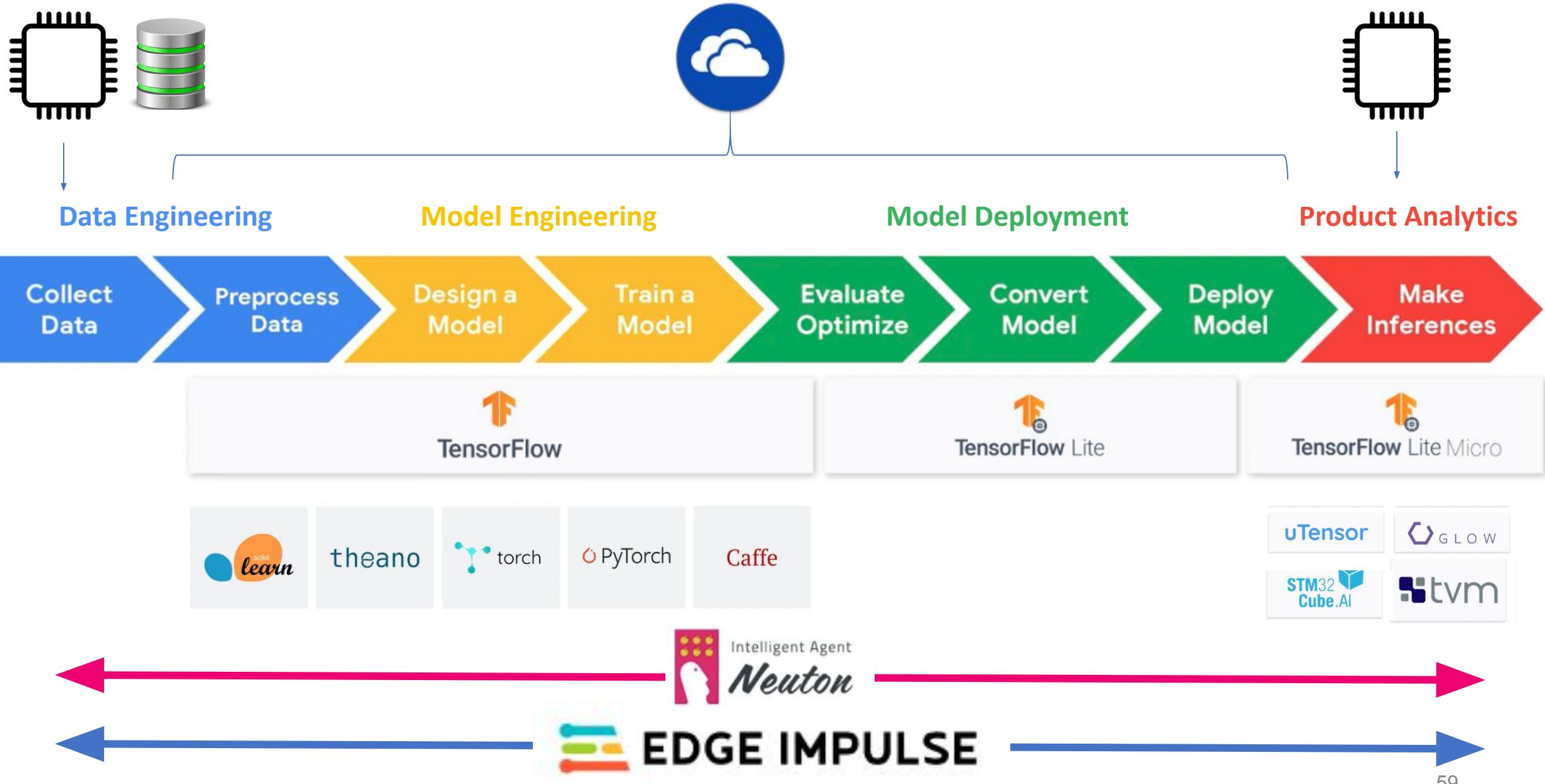












Reading Material

Main references

- [Harvard School of Engineering and Applied Sciences - CS249r: Tiny Machine Learning](#)
- [Professional Certificate in Tiny Machine Learning \(TinyML\) – edX/Harvard](#)
- [Introduction to Embedded Machine Learning - Coursera/Edge Impulse](#)
- [Computer Vision with Embedded Machine Learning - Coursera/Edge Impulse](#)
- Fundamentals textbook: “[Deep Learning with Python](#)” by François Chollet
- Applications & Deploy textbook: “[TinyML](#)” by Pete Warden, Daniel Situnayake
- Deploy textbook “[TinyML Cookbook](#)” by Gian Marco Iodice

I want to thank **Shawn Hymel** and **Edge Impulse**, **Pete Warden** and **Laurence Moroney** from Google, Professor **Vijay Janapa Reddi** and **Brian Plancher** from Harvard, and the rest of the **TinyMLEdu** team for preparing the excellent material on TinyML that is the basis of this course at UNIFEI.

The IESTI01 course is part of the [TinyML4D](#), an initiative to make TinyML education available to everyone globally.

Thanks



UNIFEI