

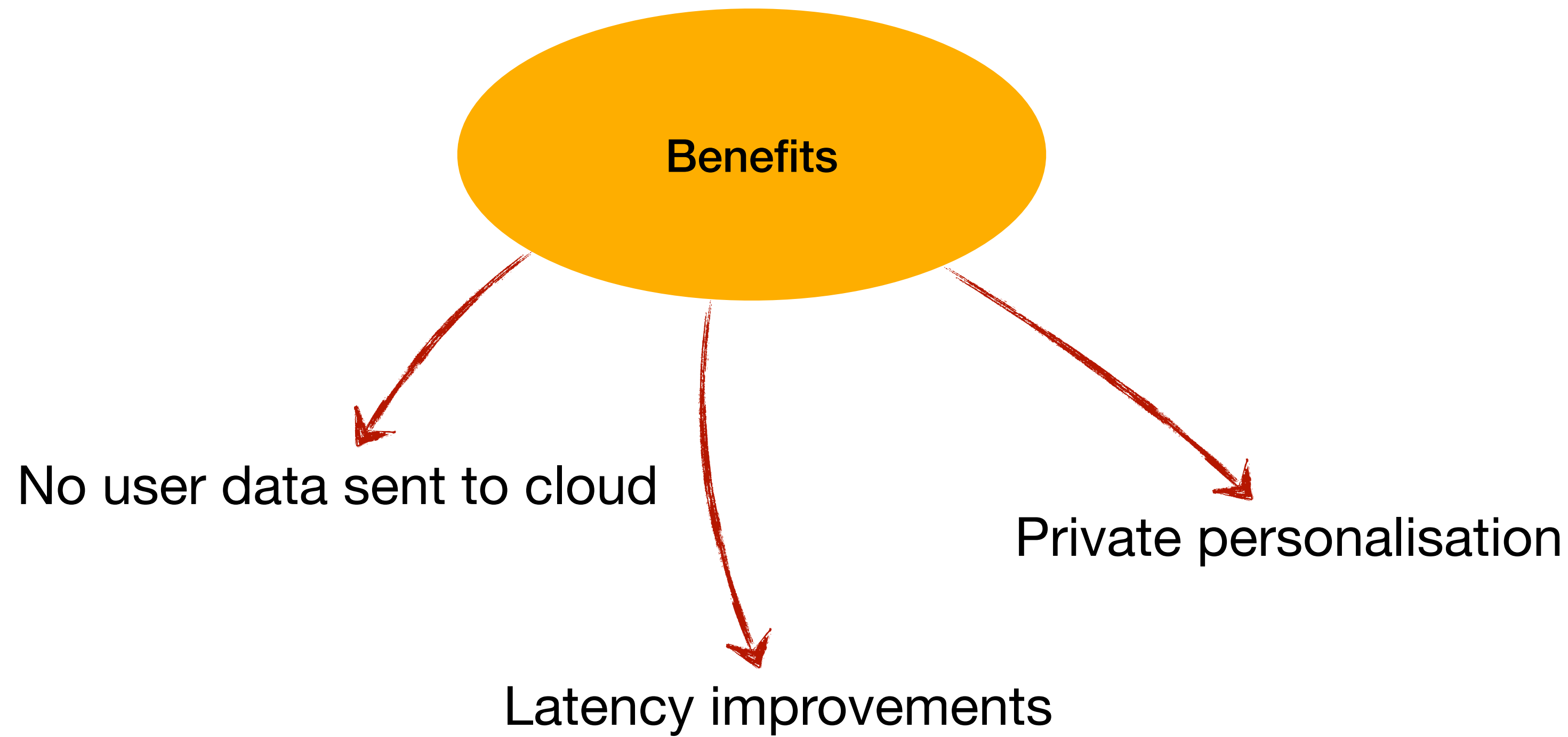






# GuaranTEE: Towards private and attestable ML with CCA

**Sandra Siby**, Sina Abdollahi, Mohammad Maheri, Marios Kogias, Hamed Haddadi

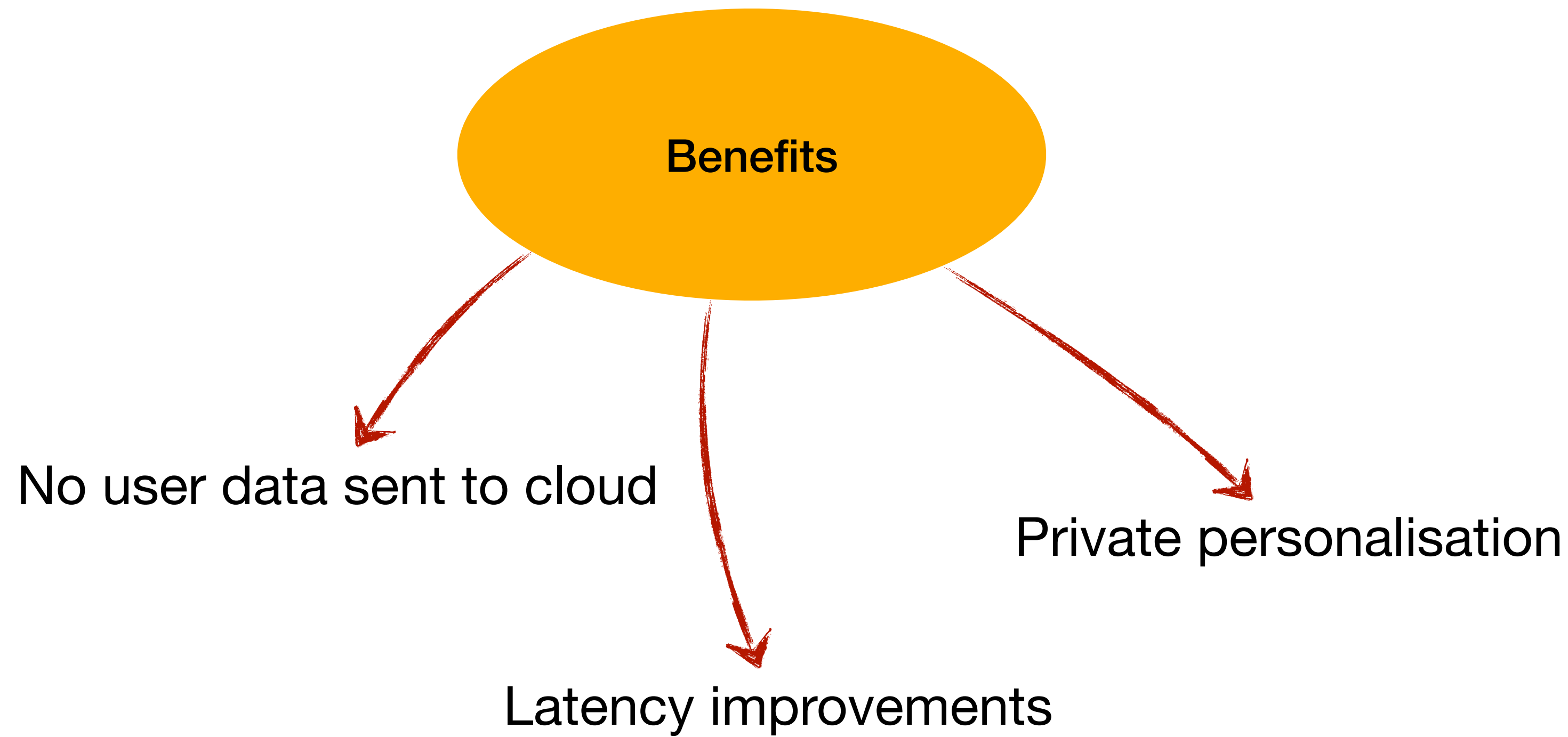
EuroMLSys, 22 April 2024

# On-device Machine Learning







 <b>Vision</b> Build features that can process and analyze images and video using computer vision.	 <b>Natural Language</b> Process and make sense of text in different ways, like embedding or classifying words.
 <b>Speech</b> Take advantage of speech recognition and saliency features for a variety of languages.	 <b>Sound</b> Analyze audio and recognize it as a particular type, such as laughter or applause.

# On-device Machine Learning

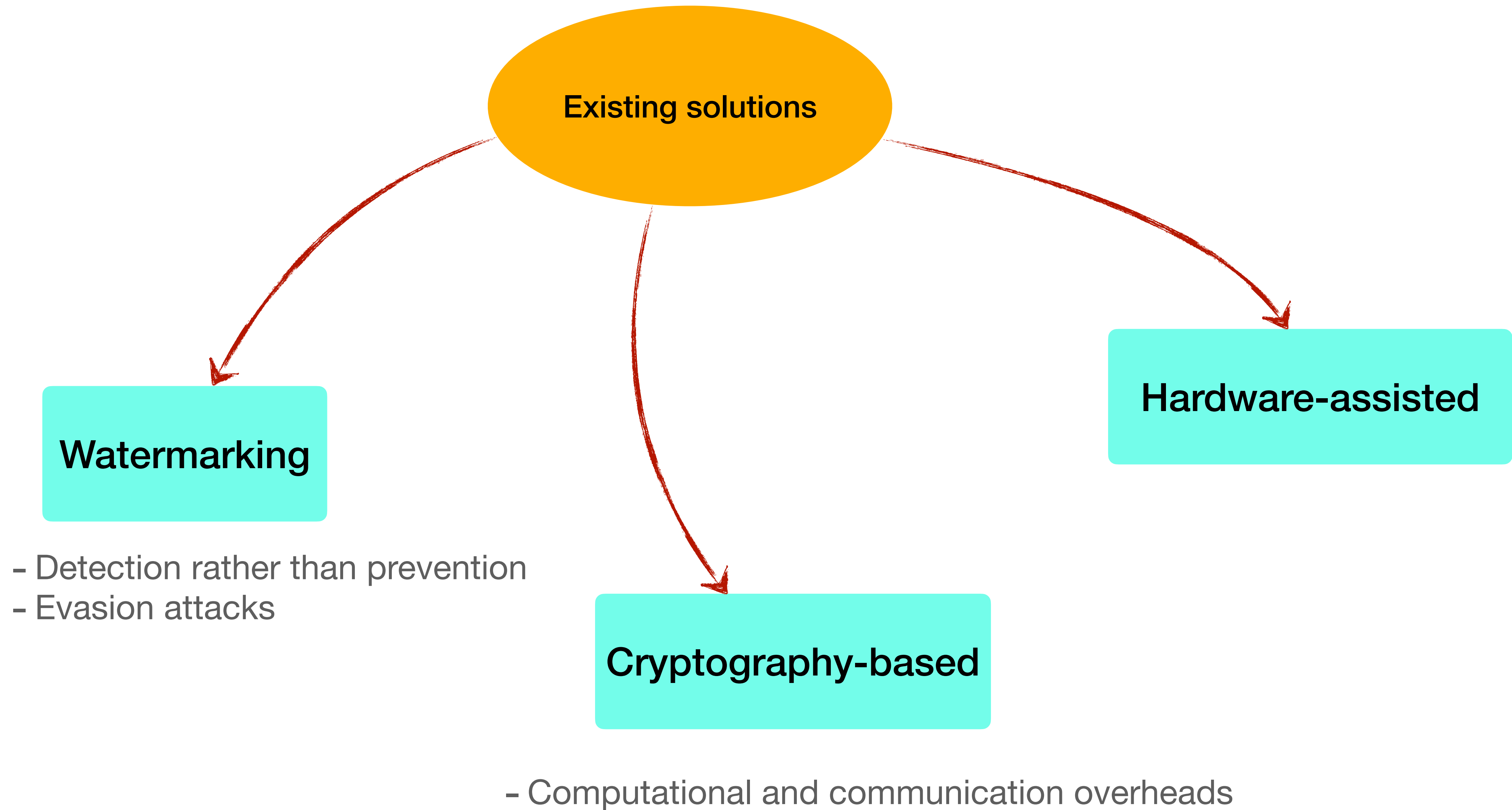


**Model providers want:**

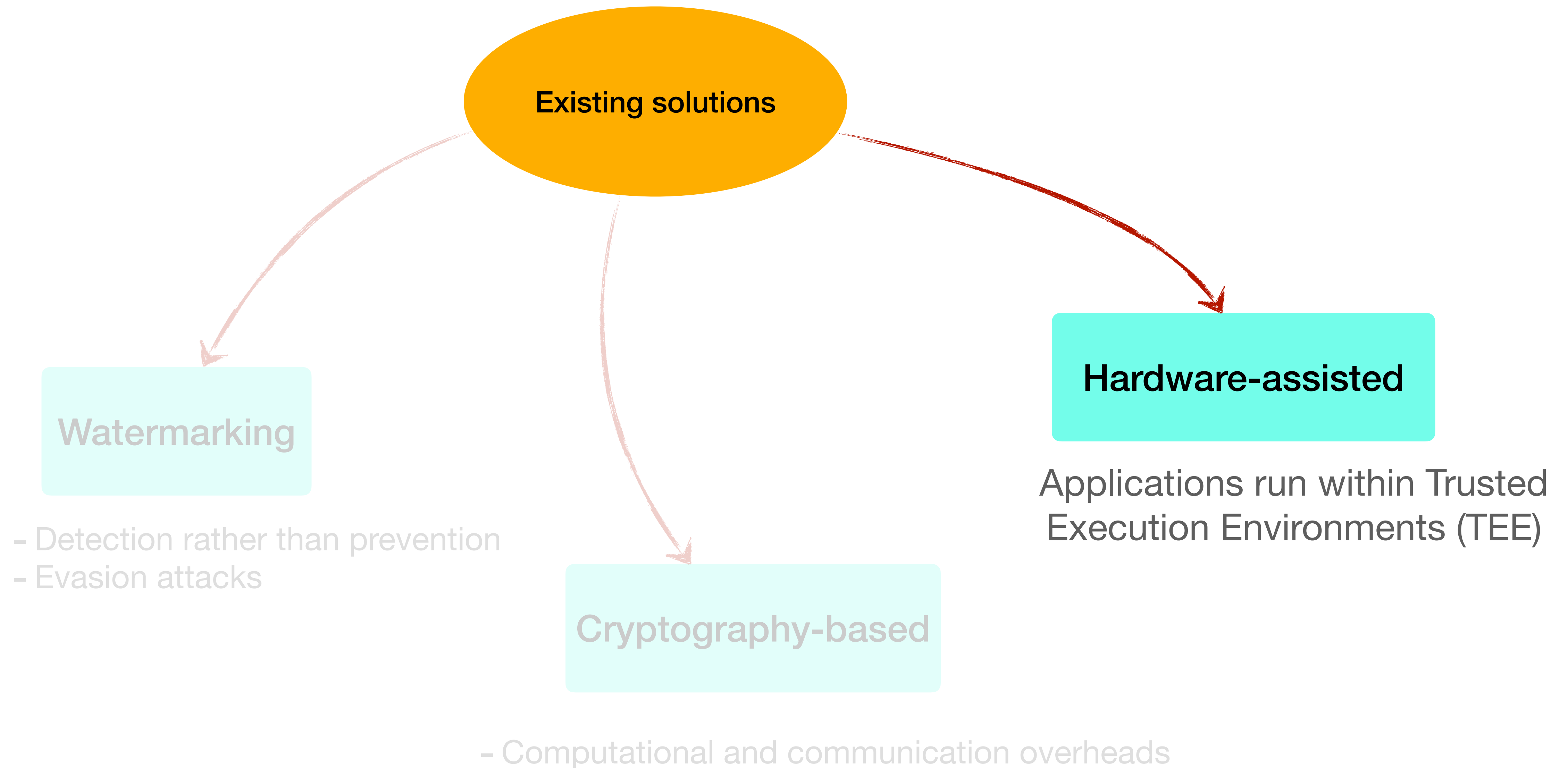
- Model privacy
- Model verifiability and attestability

 <b>Vision</b> Build features that can process and analyze images and video using computer vision.	 <b>Natural Language</b> Process and make sense of text in different ways, like embedding or classifying words.
 <b>Speech</b> Take advantage of speech recognition and saliency features for a variety of languages.	 <b>Sound</b> Analyze audio and recognize it as a particular type, such as laughter or applause.

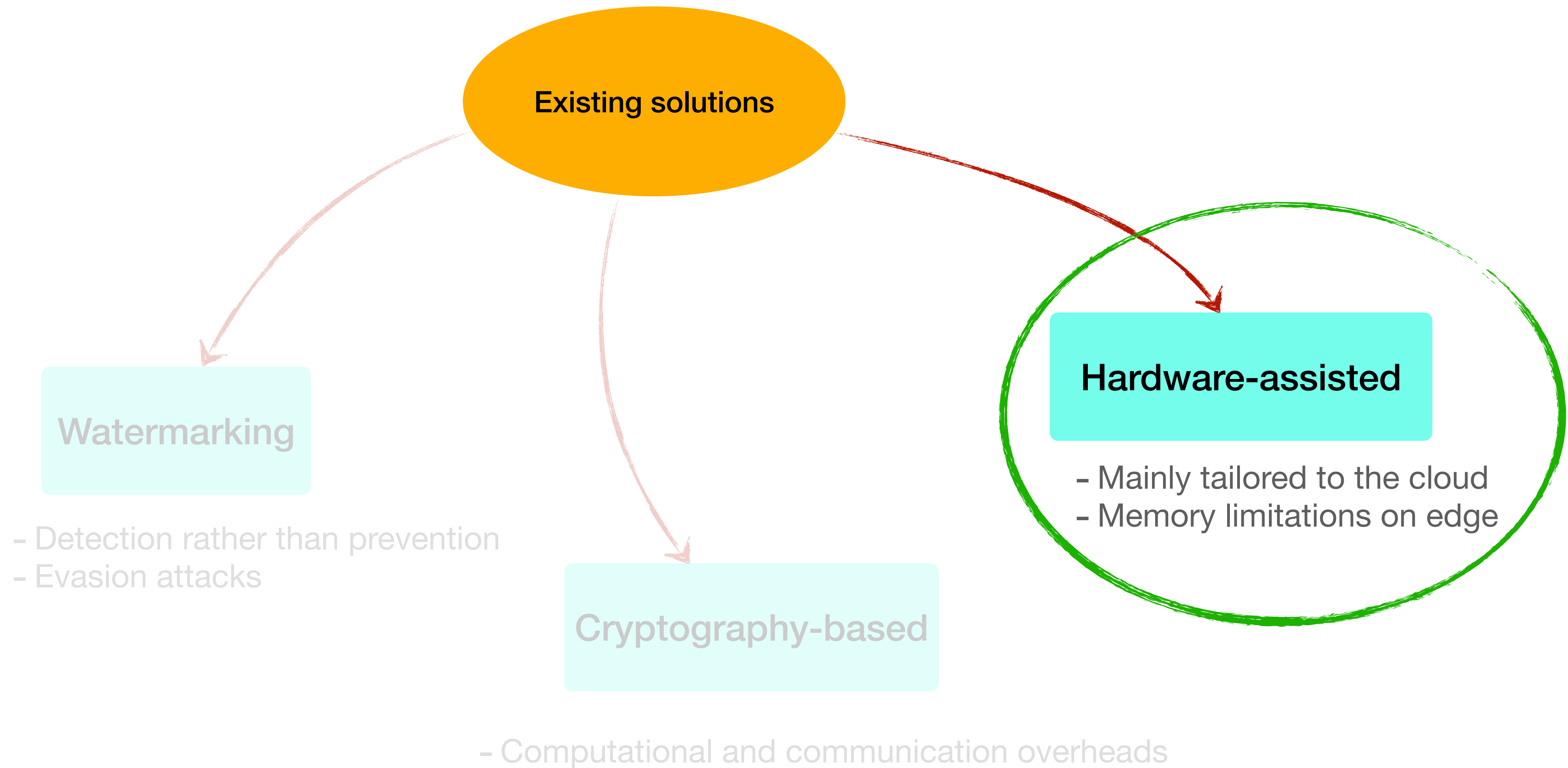
# Protecting ML models



# Protecting ML models



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# Protecting ML models

Existing solutions

## Arm's TEE solutions

Arm's TrustZone is widely deployed on edge devices.

We consider Arm's next generation of TEE solutions  
(deployment expected in 2028):

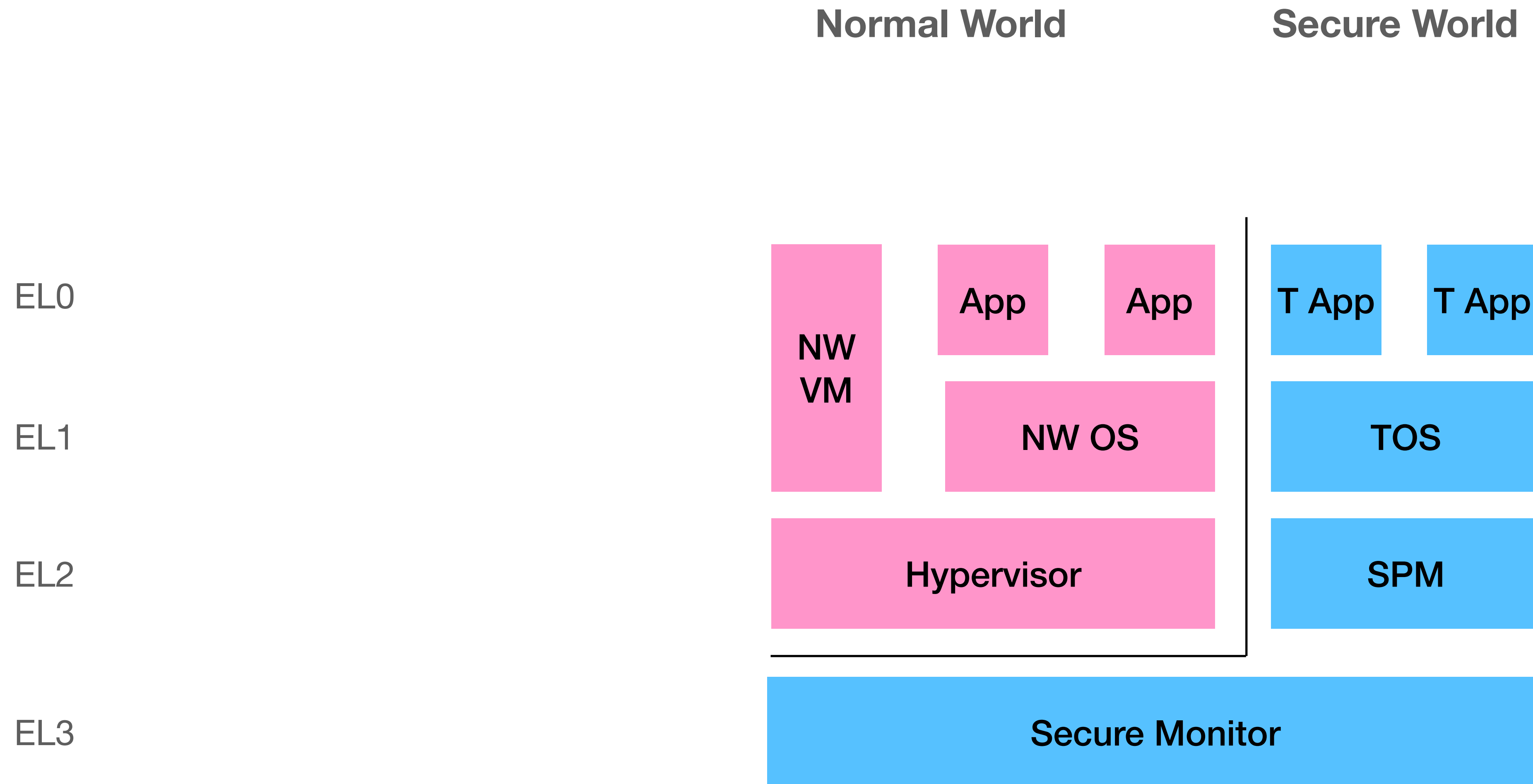
**Confidential Computing Architecture (CCA)**

## Hardware-assisted

- Mainly tailored to the cloud
- Memory limitations on edge

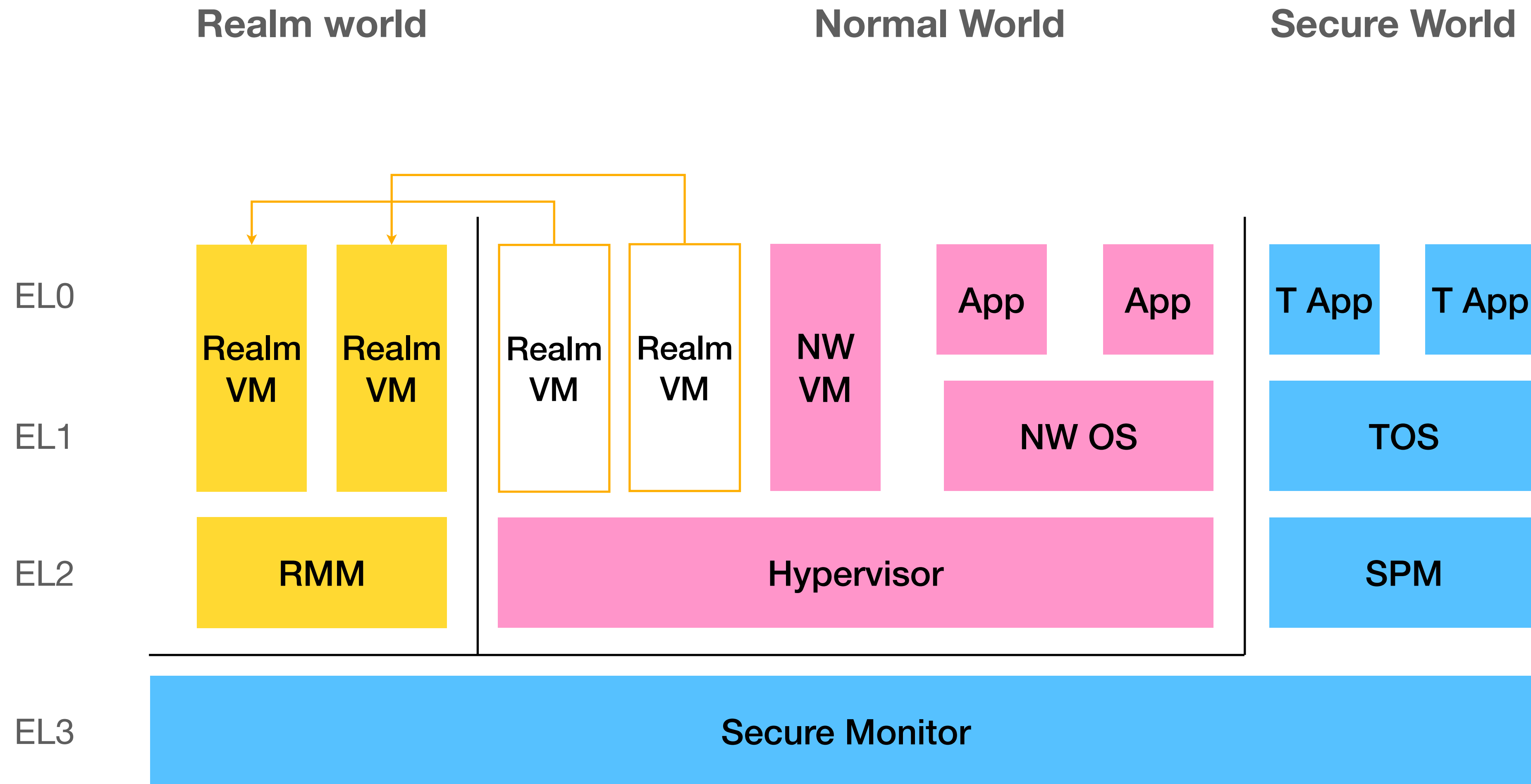
Execution overheads

# Arm TrustZone

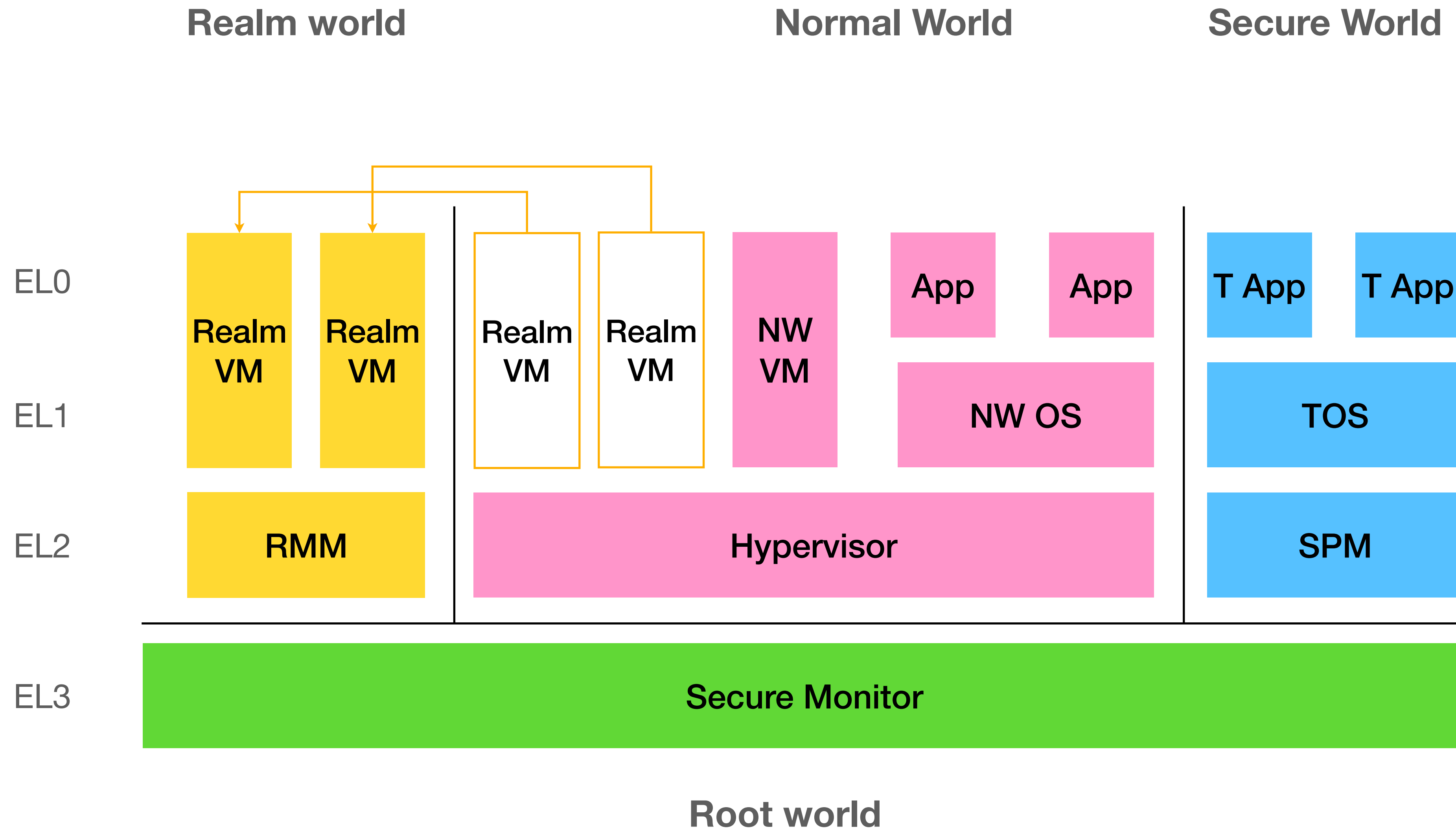




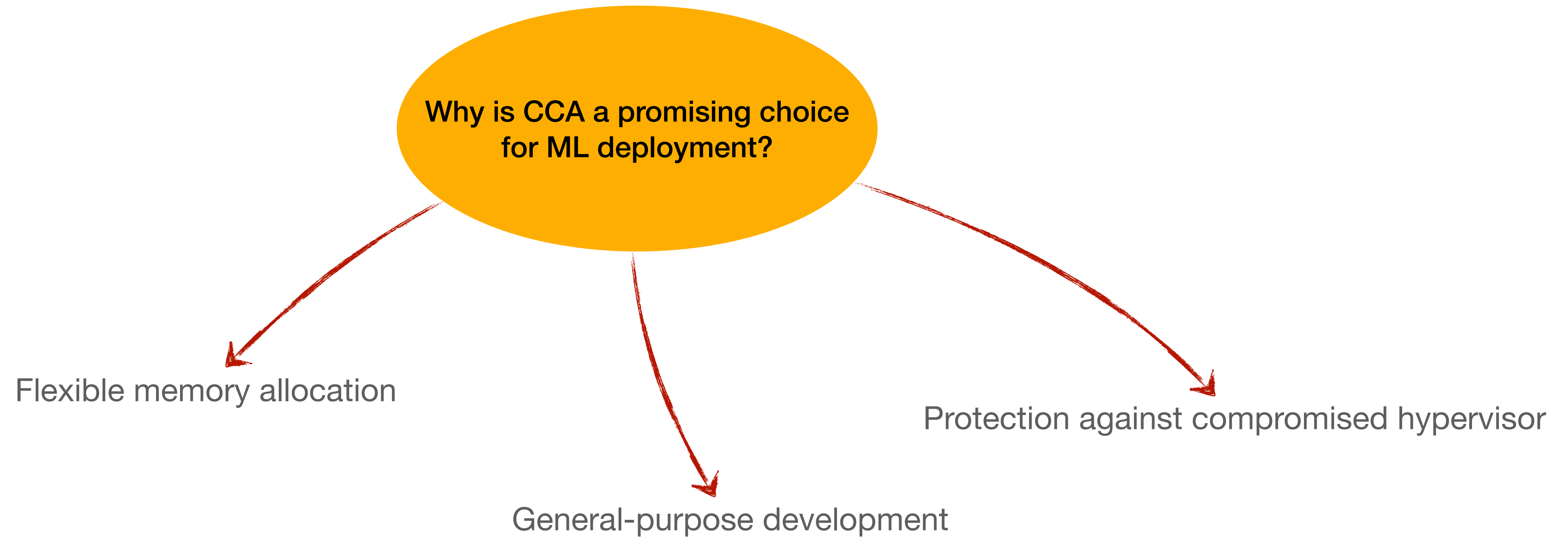
# Arm CCA



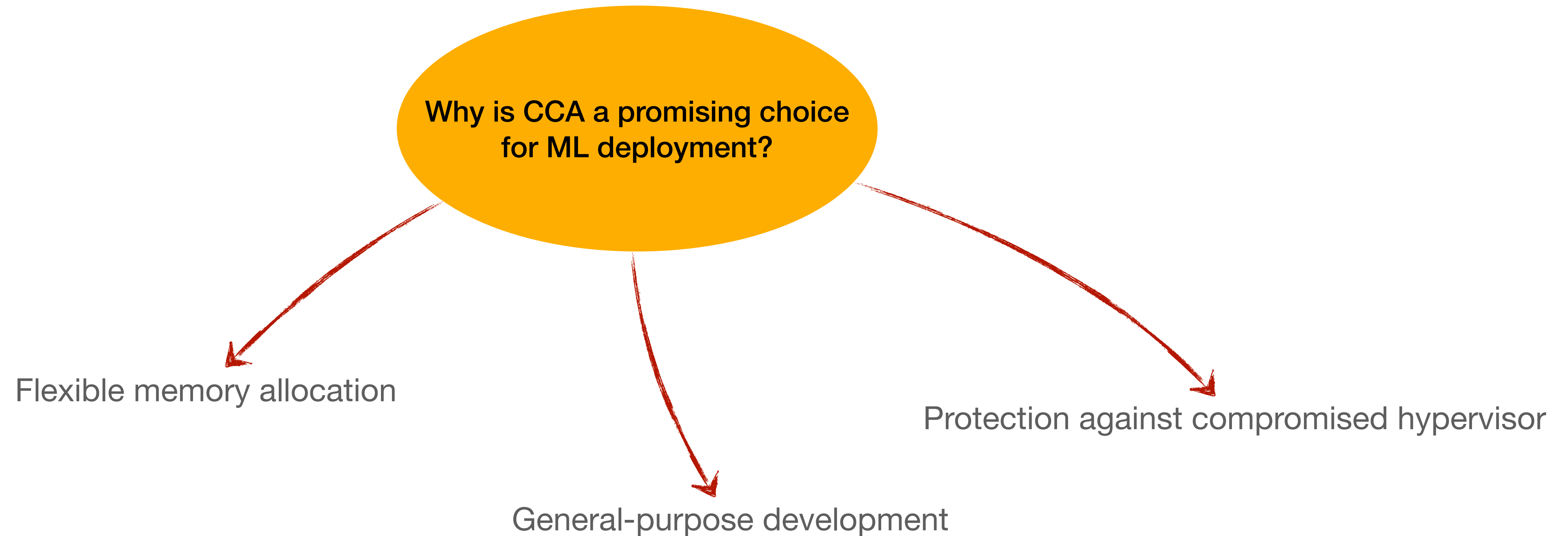
# Arm CCA



# CCA and ML deployment



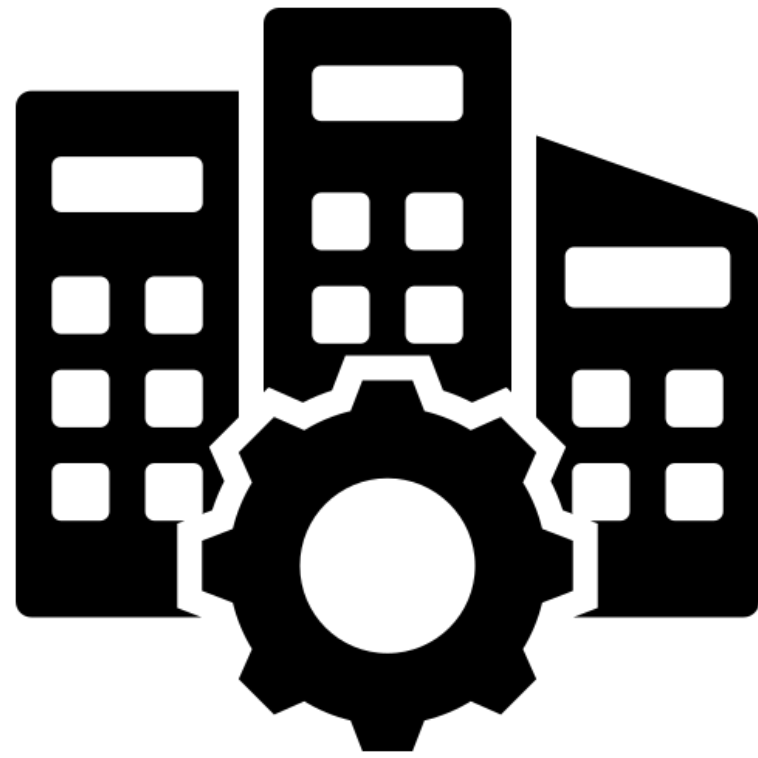
# CCA and ML deployment



## GuaranTEE

Framework for ML models to be run on end devices in a private and verifiable manner

# System overview



Model provider

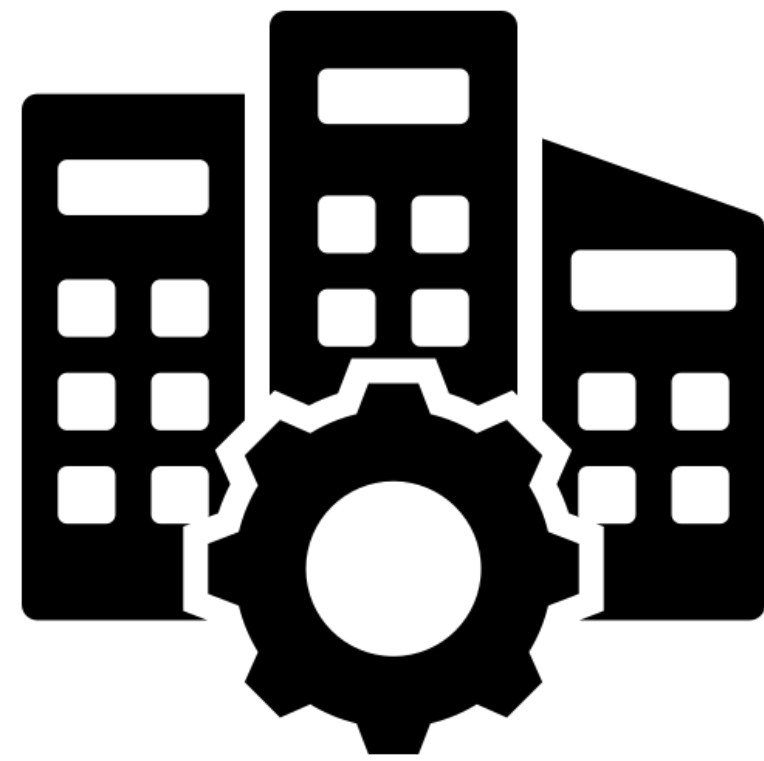


Client (Device)



Trusted verifier

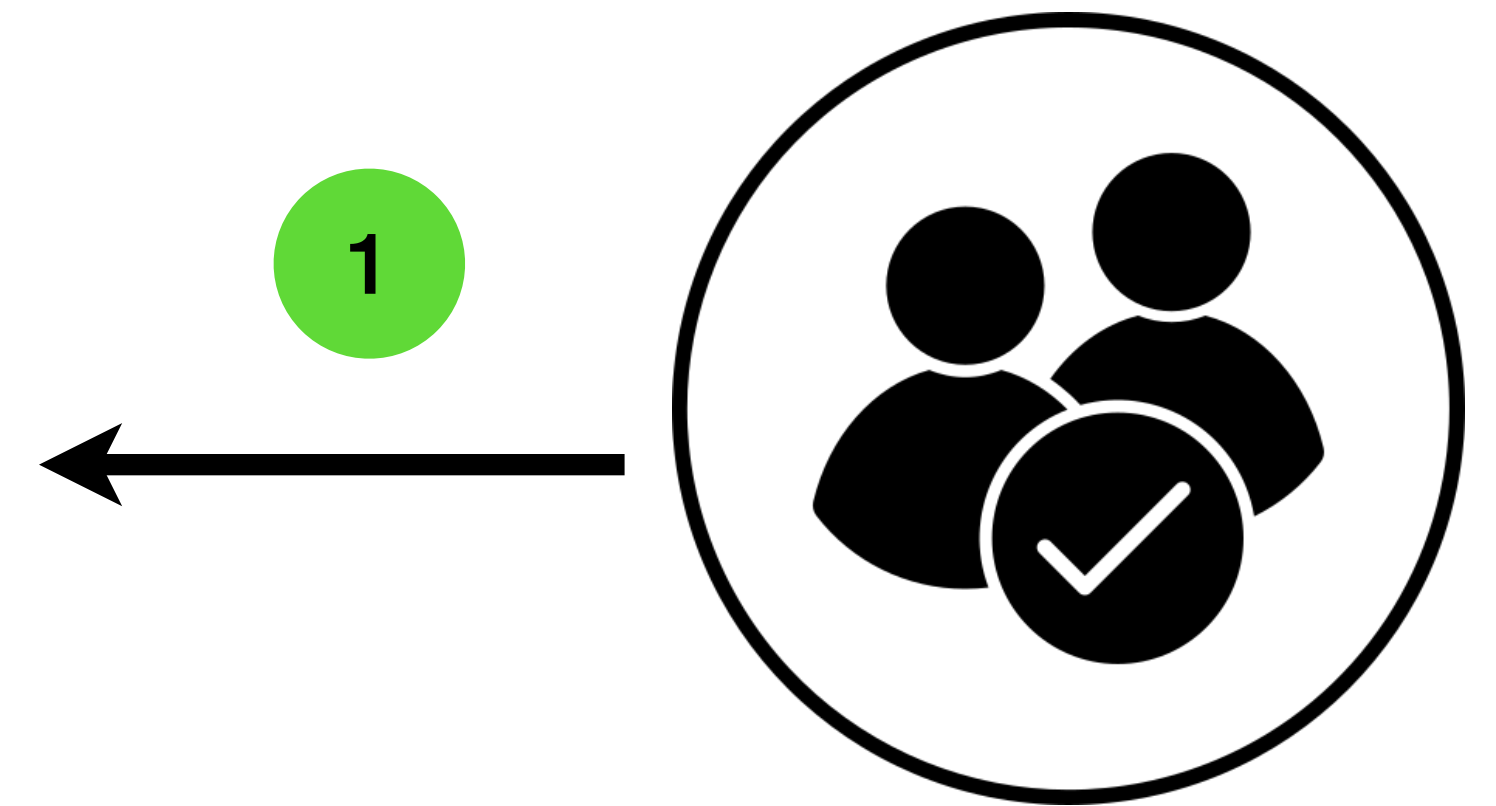
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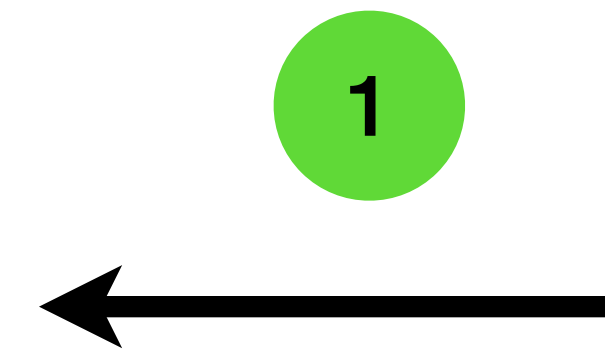
Model provider



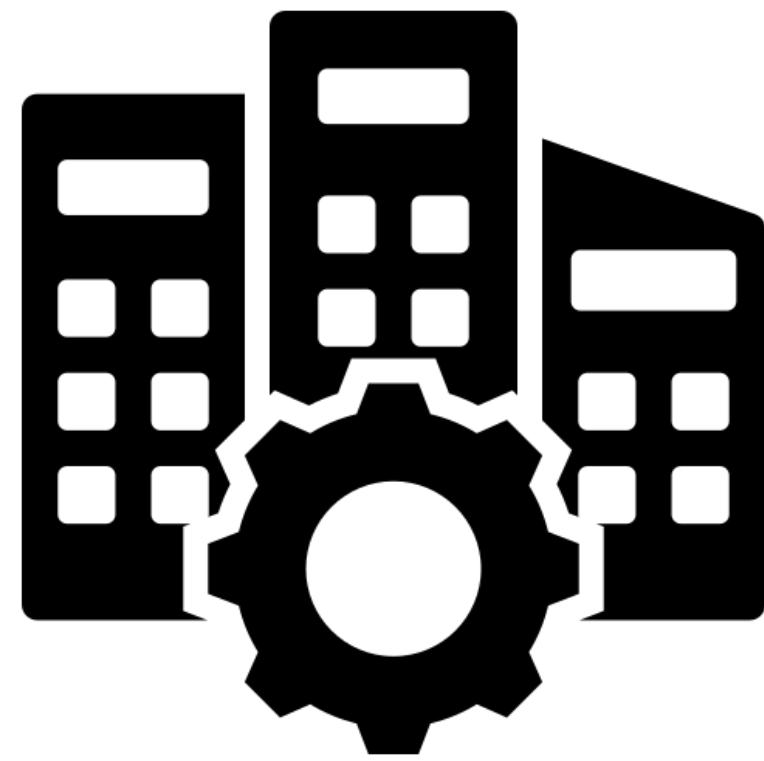
Client (Device)



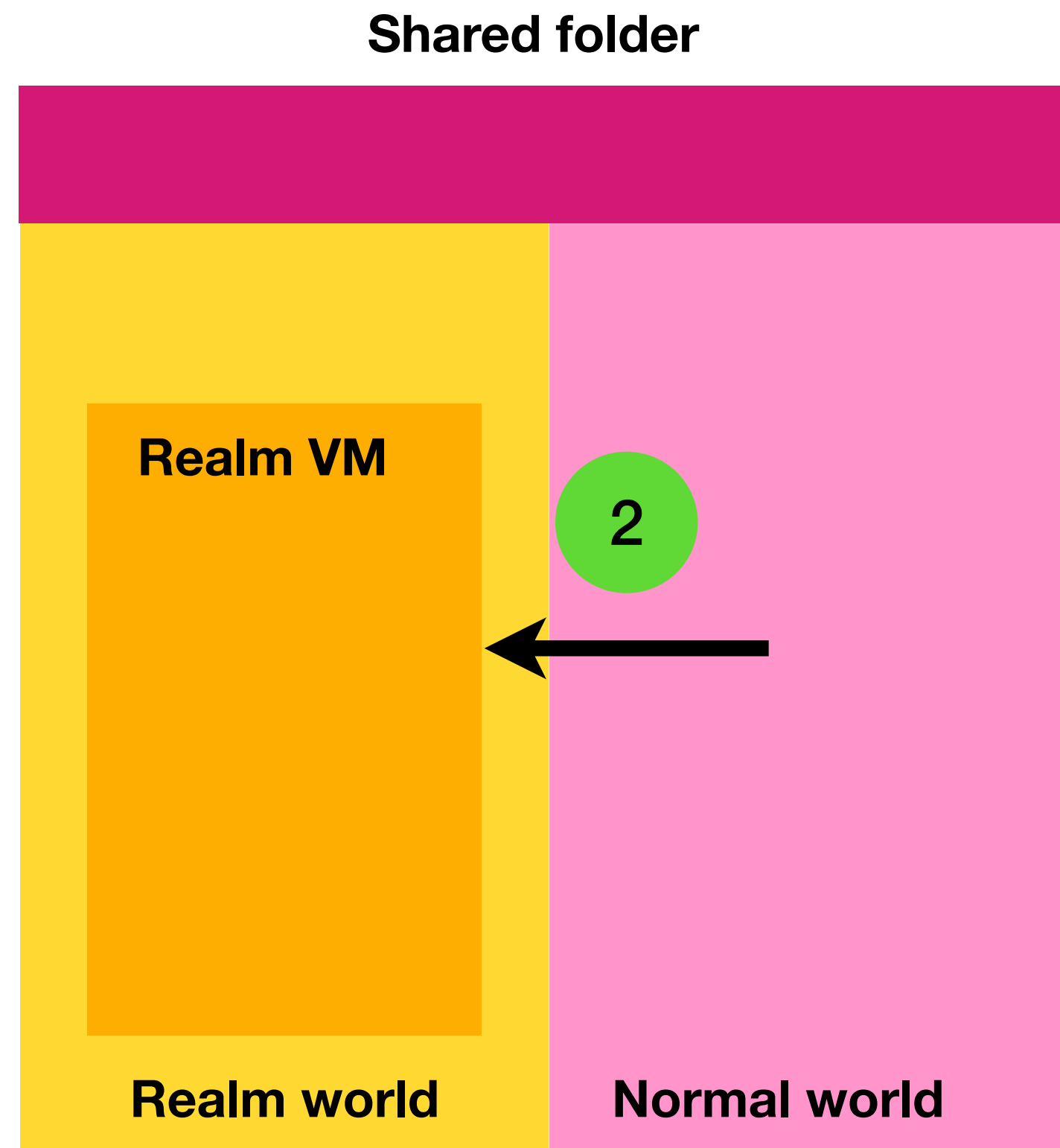
Trusted verifier



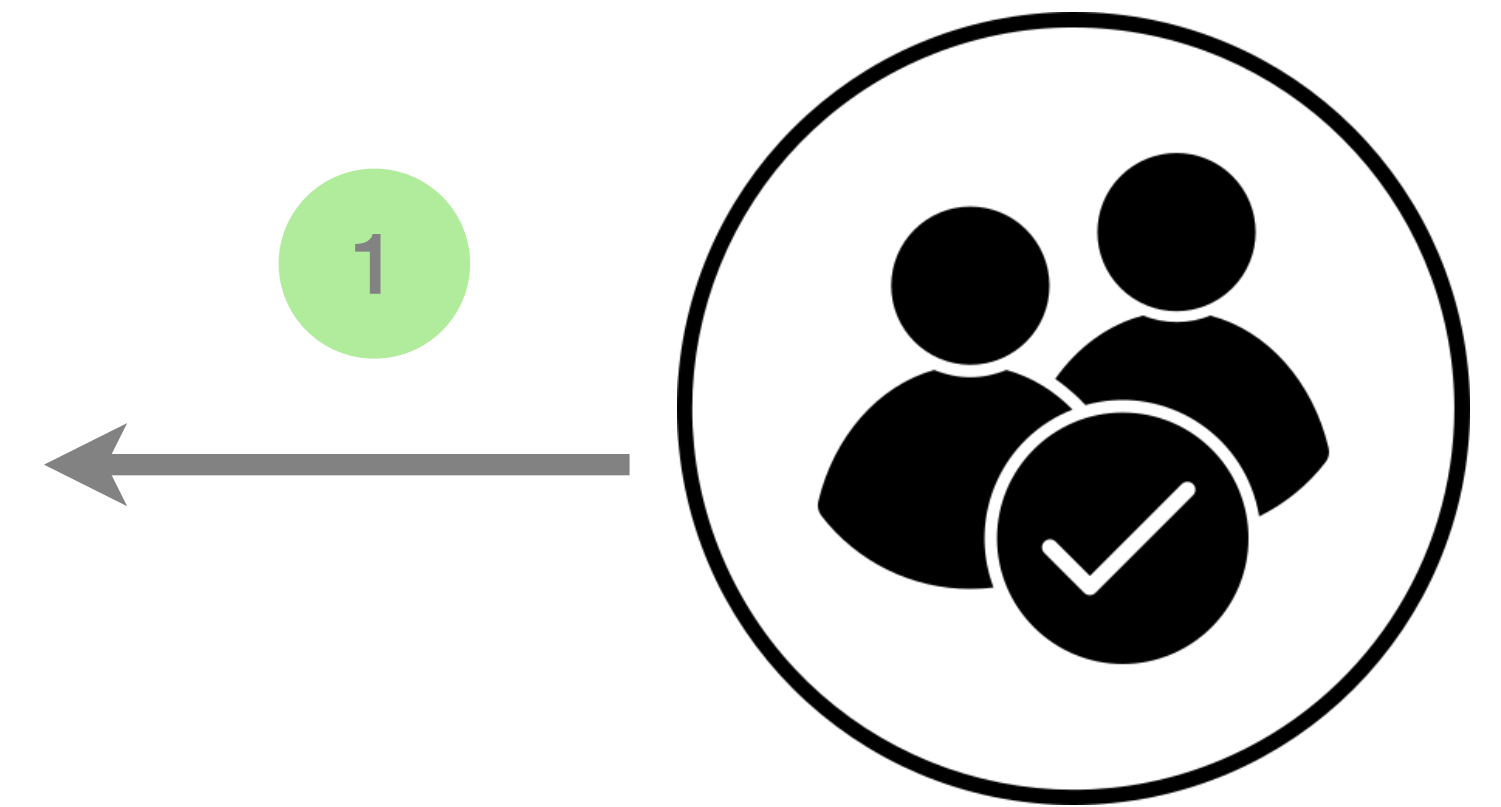
# System overview



Model provider



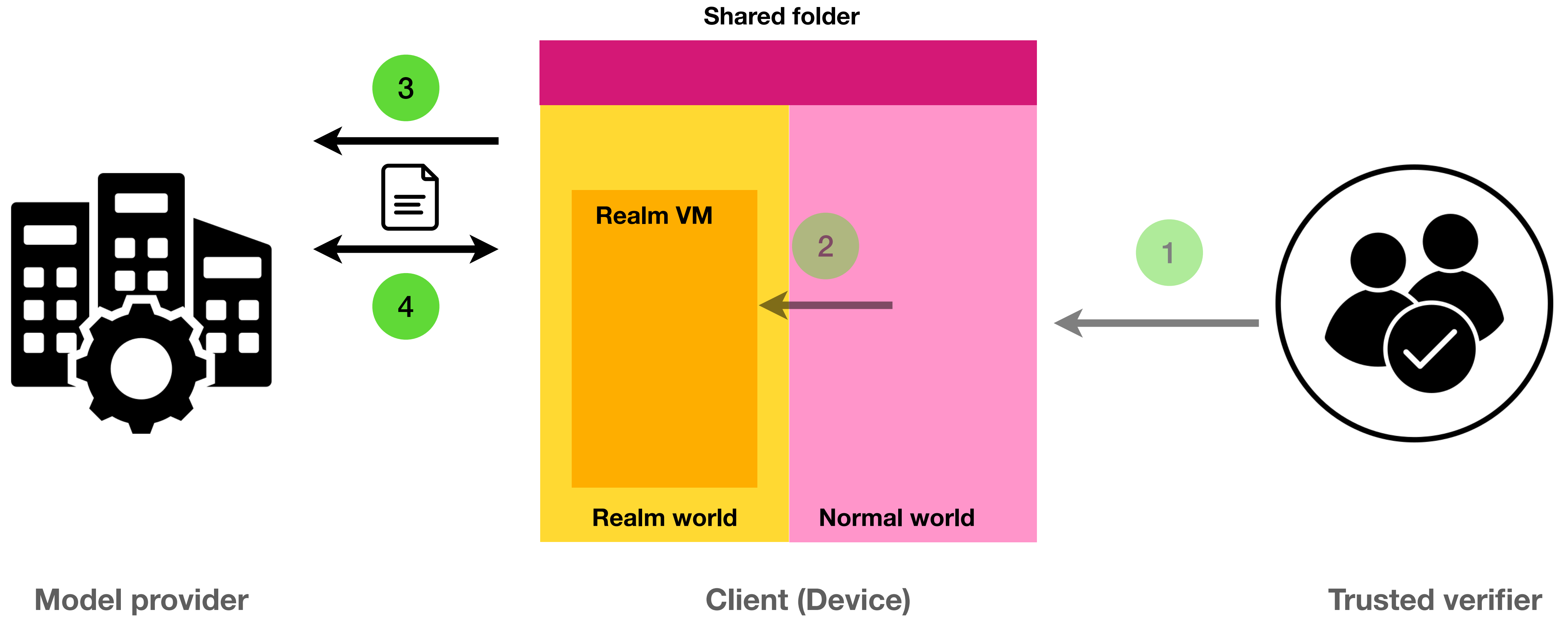
Client (Device)



Trusted verifier

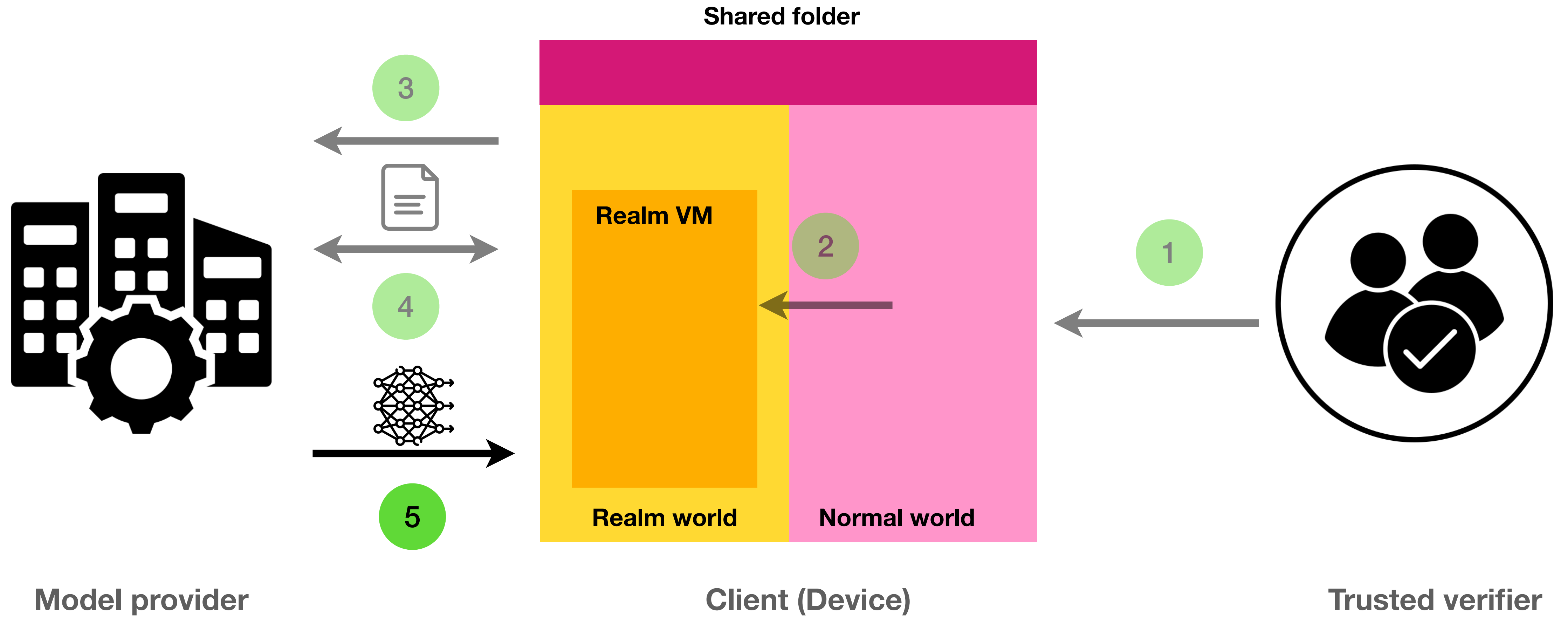


# System overview

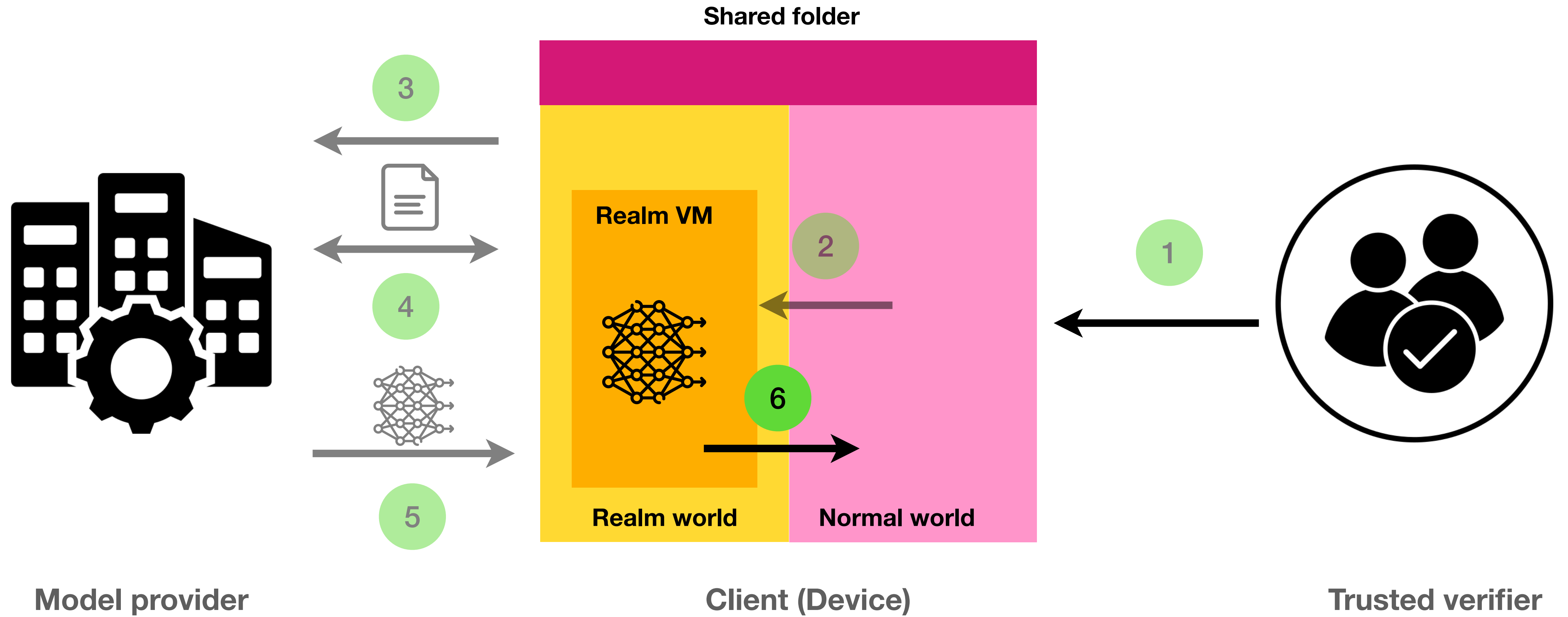




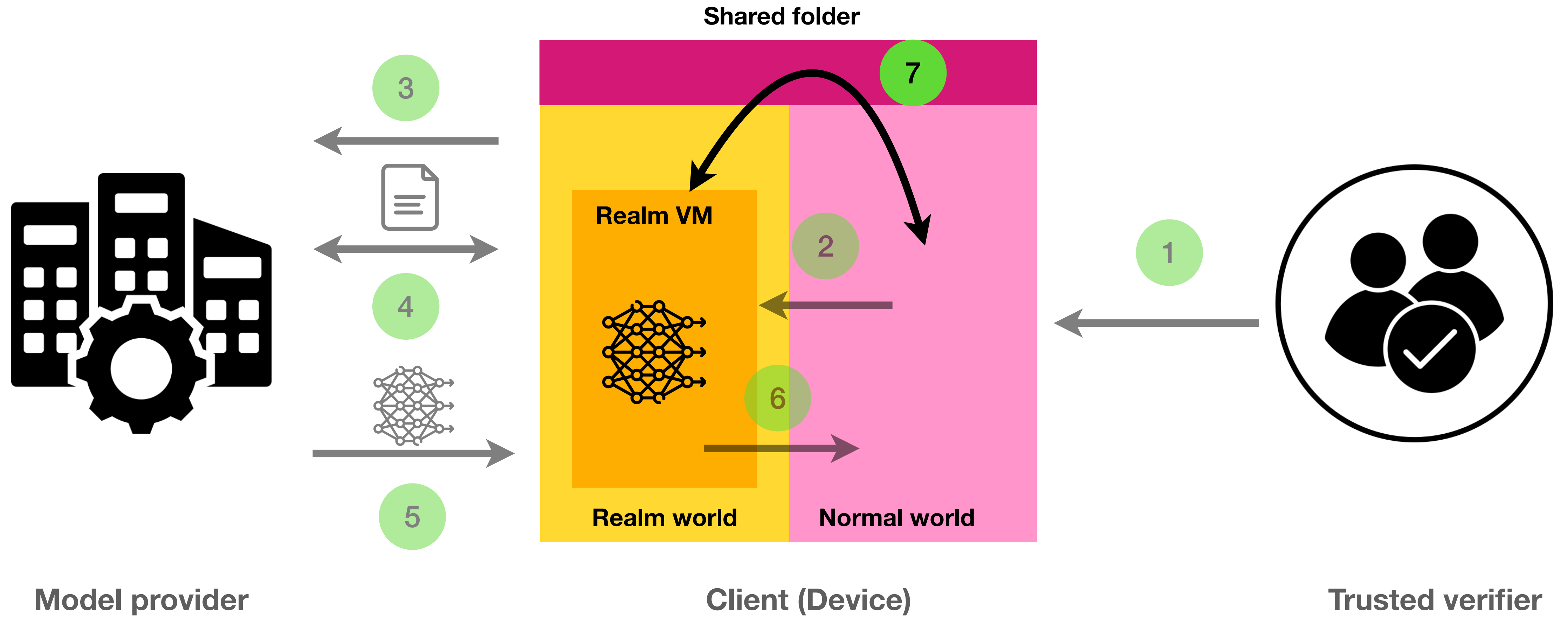
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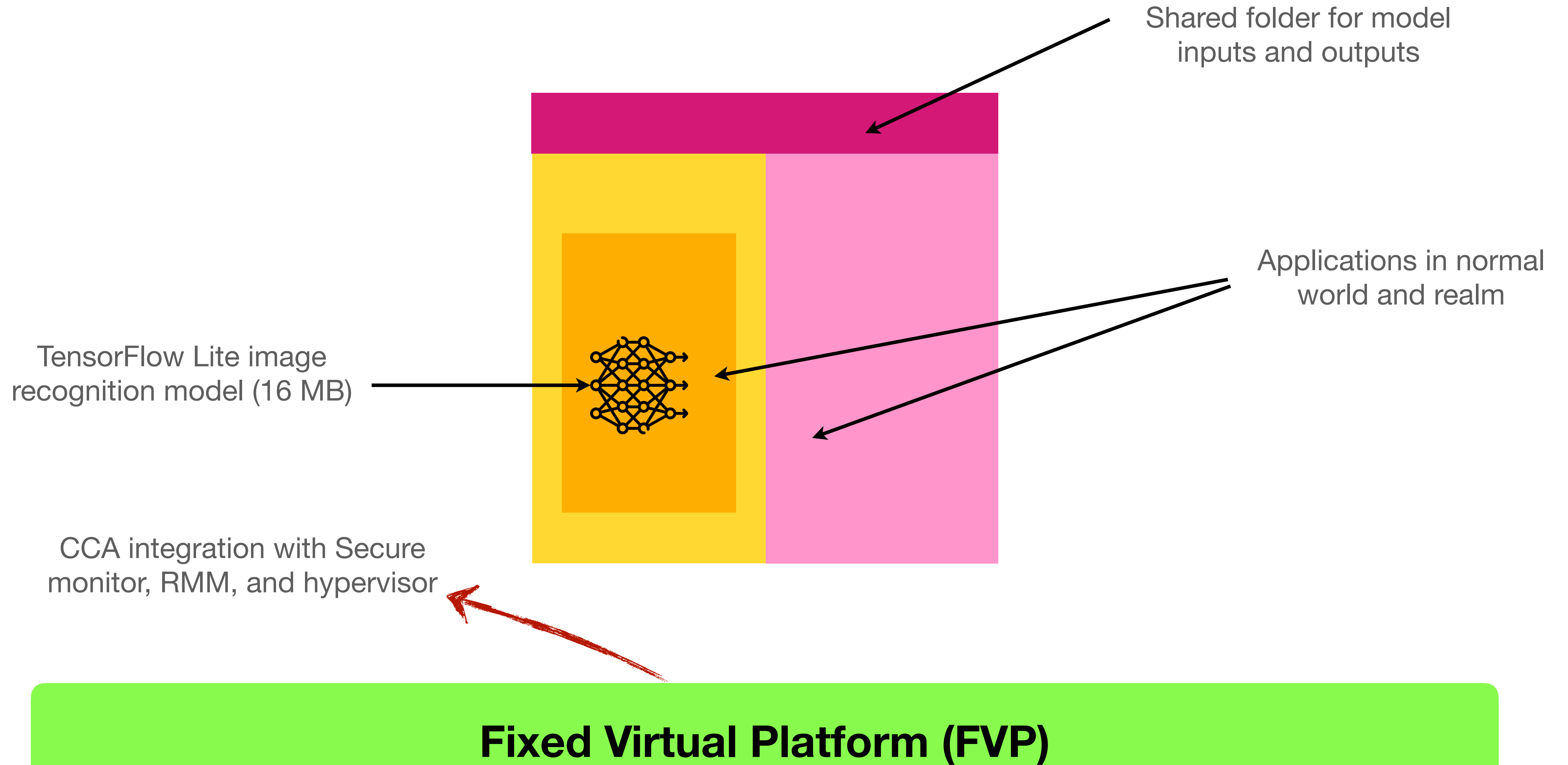
# System overview



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# Implementation



# Preliminary evaluation

**What we measure:** Overhead of inference and realm VM creation over a normal world VM.

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## Main findings

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  - Larger number of context switches
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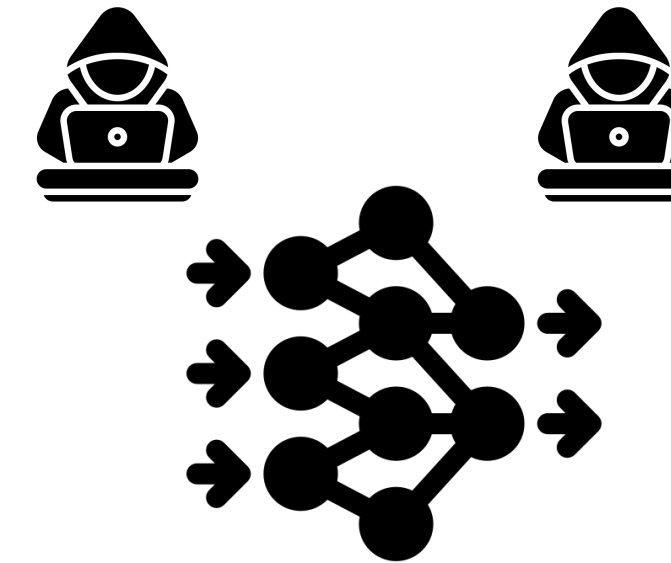
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**Note:** Full attestation report could not be implemented due to FVP limitations

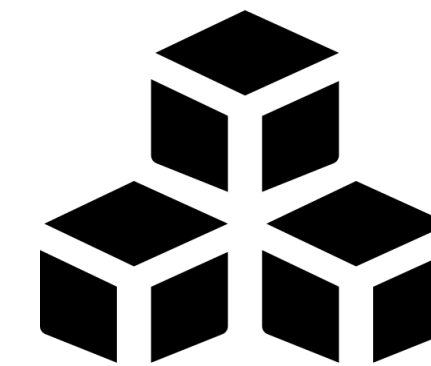


# Considerations for ML deployment with CCA

Attacks to data pipeline



Multiple providers on the same device



Policy enforcement



Availability guarantees



# Summary

- We propose GuaranTEE — a framework using CCA to deploy ML models on end devices in a private and trusted manner.
- We implement GuaranTEE using FVP, and perform a preliminary evaluation.
- We provide future directions and recommendations on ML deployment with CCA.

Code (with a setup guide): <https://github.com/comet-cc/GuaranTEE>

Get in touch: [s.siby@imperial.ac.uk](mailto:s.siby@imperial.ac.uk) ✉

