



Bringing Edge Al to





Ali Osman Örs – NXP Semiconductors David Steele – Arcturus July 13, 2021

Presenters





Ali Osman Örs Director, Al ML Strategy and Technologies NXP Semiconductors





David Steele
Director of Innovation
Arcturus

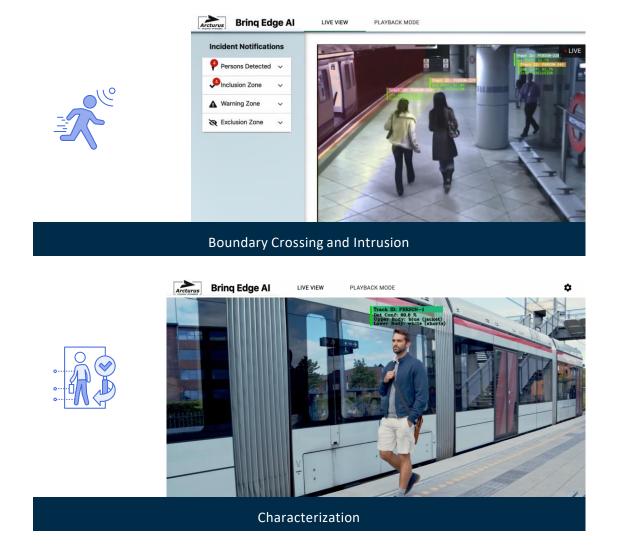


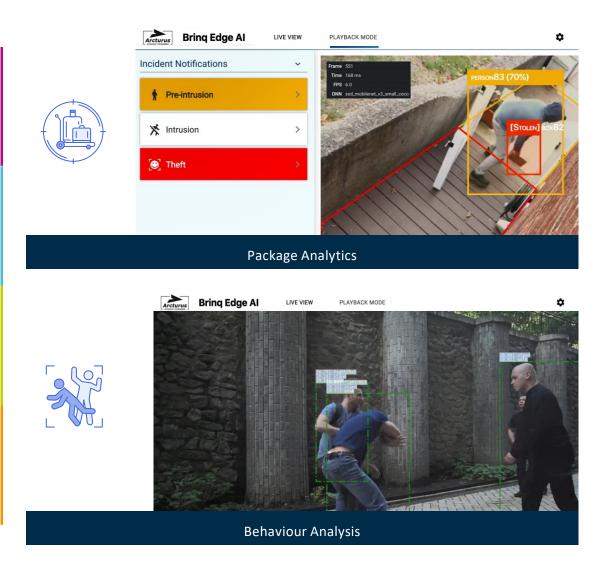
Agenda

- Introduction
- Tools and Enablement
- Building and Scaling Applications
- Optimization Techniques
- Demo
- 2x \$250 Amazon Gift Card Draw



Introduction to Edge AI Applications







Development Challenges

- Development paradigm
 - Experimentation and adaptation
 - Heavily reliant on optimization
 - Complex interdependencies
 - Requires broad expertise
- Starting points
 - Do not move to production seamlessly
 - Dependencies / edge runtime limitations
 - Lab vs field data
 - Missing components / specialized requirements
- Considerations
 - Move an application to the edge
 - Improve accuracy and performance
 - Develop a scalable/flexible architecture





Edge Enablement and Tools

Ali Osman ÖrsNXP Semiconductors

NXP Broad-based Machine Learning Solutions and Support



elQ Machine Learning SW

elQ™ ML SW Development Environment

eIQ Toolkit with eIQ Portal GUI to:

- Import/create, convert, optimize, validate and deploy ML models
- Dataset curation tools to create new, augment, label/annotate datasets

elQ inference with: TensorFlow Lite, TensorFlow Lite Micro, Arm NN, ONNX Runtime, Glow and DeepViewRT

Support for i.MX 8 family, i.MX RT family

Integrated into NXP development environments (MCUXpresso, Yocto/Linux)



elQ Auto

elQ™ Auto Al Enablement

Deep Learning toolkit for S32V processors

Auto Quality: A-SPICE qualified inference engine

Optimization: Prunes, quantizes, compresses the Neural Network

Automated neural net layer deployment to optimum available compute resource

Automotive Grade



CORAL

Third Party SW and HW

Google Coral Dev Board

DIY

i.MX 8M Mini Development Kit for Amazon® Alexa Voice Service

Au-Zone Value-add packages for NXP eIQ Toolkit

Arcturus video applications

SensiML tools for sensor analysis

.... And more



SLN-ALEXA-IOT

Turnkey Solutions

Alexa Voice Services (AVS) solution

• i.MX RT106A (kit – SLN-ALEXA-IOT)

Local voice control solution

• i.MX RT106L (kit - SLN-LOCAL-IOT)

Face & emotion recognition solution with Anti-Spoofing

• i.MX RT106F (kit – SLN-VIZN-IOT)

Fully Tested



NXP Scalable Edge Processing Continuum

CROSSOVER

i.MX RT

Scalable/reusable software, tools and solutions

Arm Cortex-A and Cortex-M cores APPLICATIONS PROCESSORS 532x | i.MX | Layerscape

Future i.MX 9 MPUs

Example Applications:

- Smart Cities
- Smart Homes
- **Smart Buildings**
- **Smart Factories**

Cortex-A cores & Cortex-M cores First use of Arm Ethos-U65 NPU

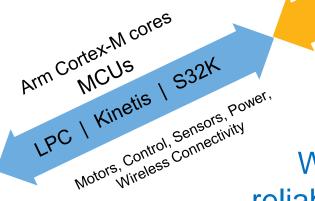
Voice | Audio Graphics

Example Applications:

- Machine vision
- Industrial computer
- Security and vision



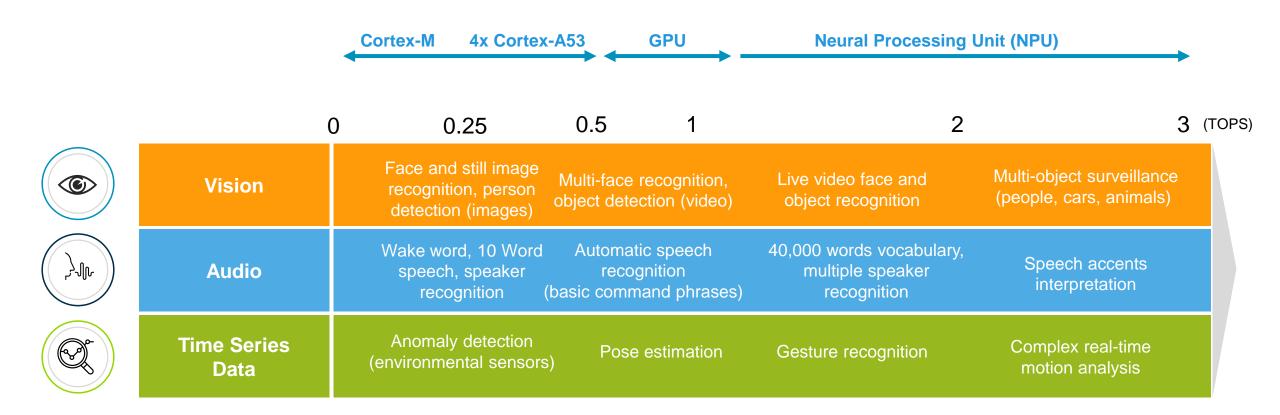
Quad/Dual Cortex-A53 and M7 2.3 TOPS NPU, GPU, DSP Camera, display, video, audio CAN-FD, GbE, TSN, PCIe



With Industrial-grade reliability, security, longevity

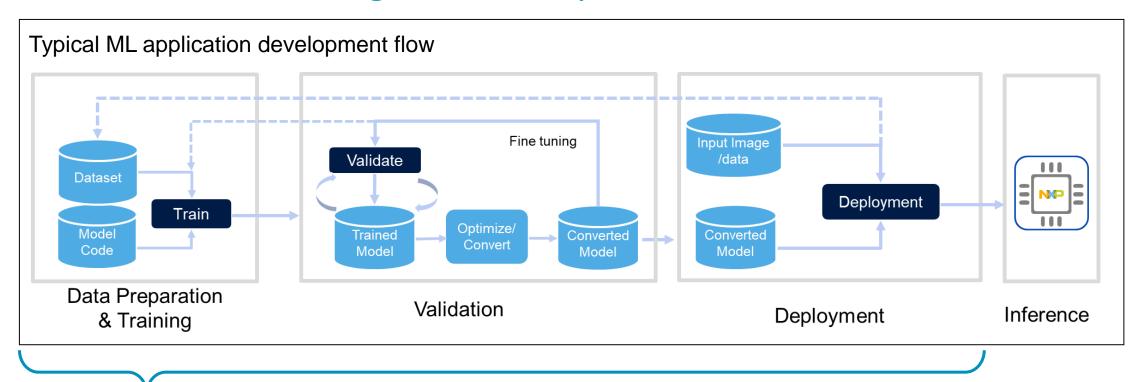


Machine learning use cases and accelerators





elQ Machine Learning SW Development Environment

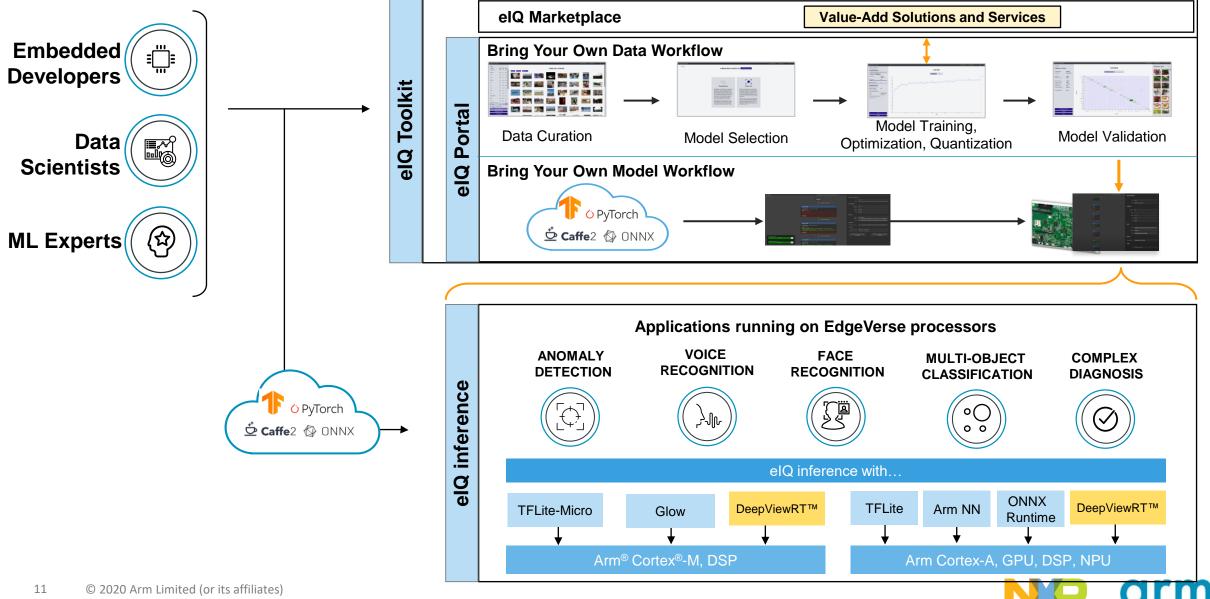


NXP's **eIQ ML Software** provides a collection of development tools, utilities and libraries for building ML applications using NXP MCUs and applications processors (MPUs).

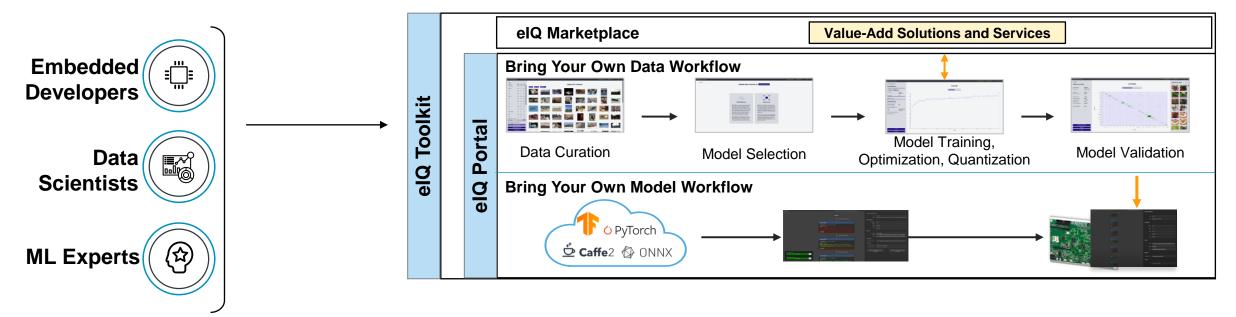
elQ ML software can be leveraged as part of a user's existing flow or can be used for the complete flow depending on the ML application targeted.



elQ[™] ML SW Development Environment



eIQ Toolkit and eIQ Portal



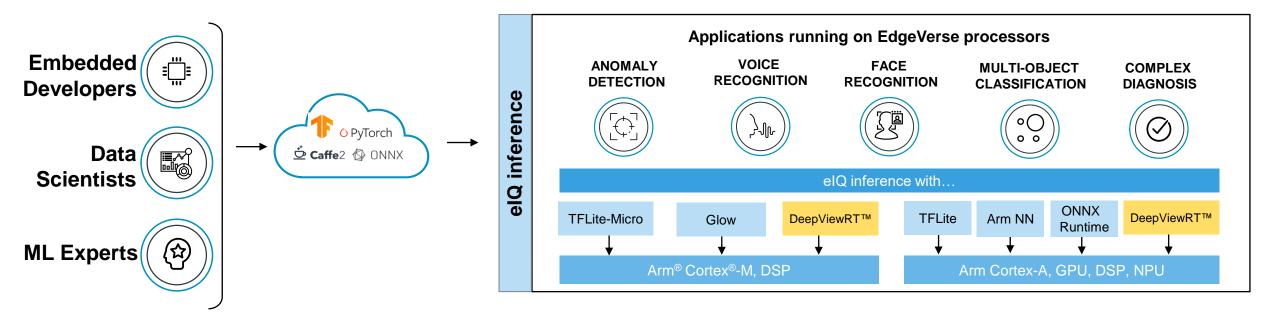
eIQ Toolkit

- Eases the ML development experience with the eIQ Portal as well as with command line host tools like; Glow tools, TensorFlow, PyTorch and other third-party tools
- Enables graph-level profiling capability with runtime insights to help optimize neural network architectures for execution on target NXP processors
- Includes ML application examples
- **eIQ Portal** intuitive graphical user interface (GUI) that simplifies ML development:
 - Creates, optimizes, debugs, converts, and exports ML models
 - Imports datasets and models, rapidly trains and deploys neural network models and ML workloads
 - Output seamlessly feeds into DeepViewRT, TensorFlow Lite, TensorFlow Lite Micro, Glow, Arm NN, and ONNX Runtime inference engines
 - Can import models from TensorFlow and PyTorch ML frameworks
 - Includes object detection and image classification models for computer vision applications
- eIQ Marketplace offers value-add solutions, professional support and design services from trusted eco-system partners and NXP
- Delivered with a single click from the eIQ Toolkit at <u>www.nxp.com/eIQ</u>





elQ inference



elQ inference with TensorFlow Lite, ARM NN, Glow and ONNX Runtime

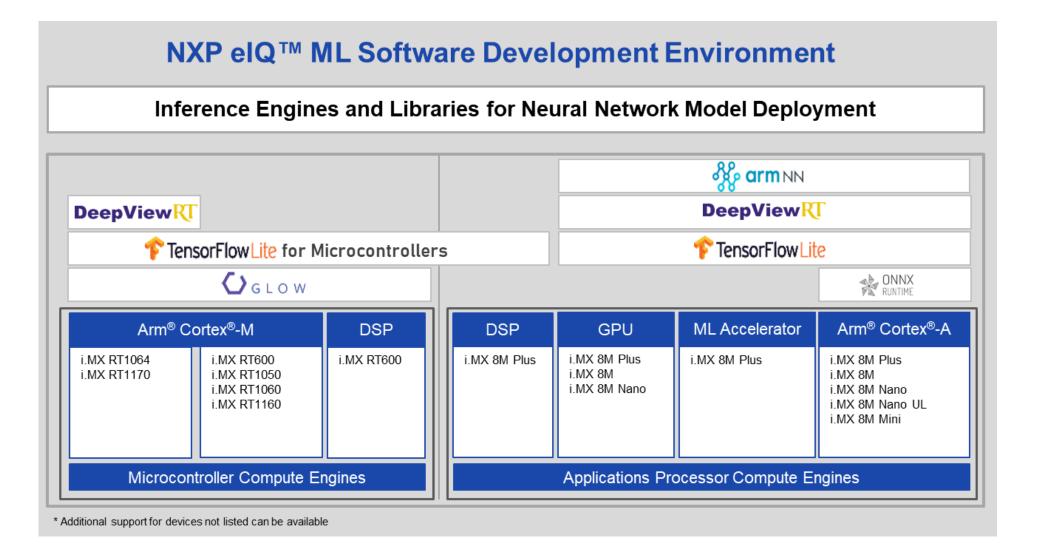
- Optimized support for open community-based engines on NXP target devices

eIQ inference with DeepViewRT runtime

- Platform-optimized, proprietary runtime inference engine that scales across a wide range of NXP devices and neural network compute engines
- Stable and longer-term maintained solution to complement the open community-based inference solutions.
- Supports EdgeVerse™ processors, including the i.MX RT crossover MCUs (Arm® Cortex®-M cores), i.MX applications processors (Cortex-A and Cortex-M cores, dedicated Neural Processing Units (NPU) and GPUs)
- Enables compact code size for resource-constrained devices with ease of analysis and fine-tuning of model performance using the eIQ Toolkit
- Delivered via NXP standard <u>Yocto BSP release</u> for Linux® OS-based development, and <u>MCUXpresso SDK</u> release for embedded RTOS-enabled MCU development
- eIQ inference with DeepViewRT runtime is provided free-of-charge to NXP customers as part of our ML enablement



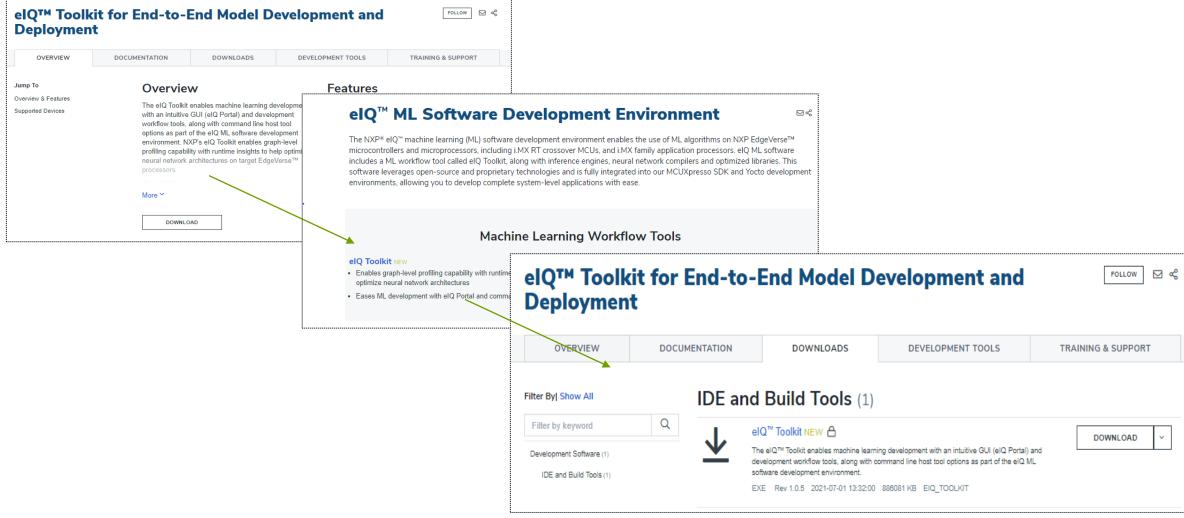
NXP eIQ[™] ML Software Development Environment





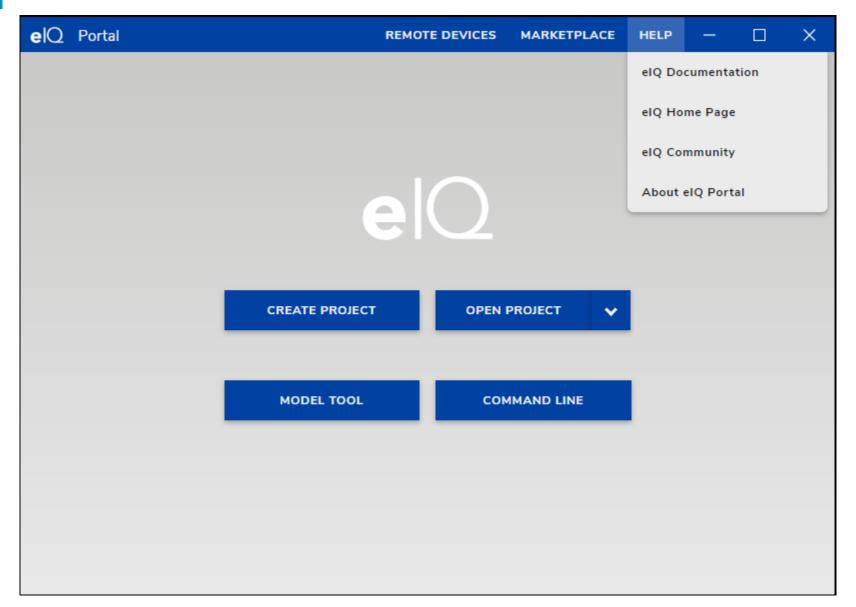
eIQ Toolkit availability

Registered users can download eIQ Toolkit from: http://www.nxp.com/eiq





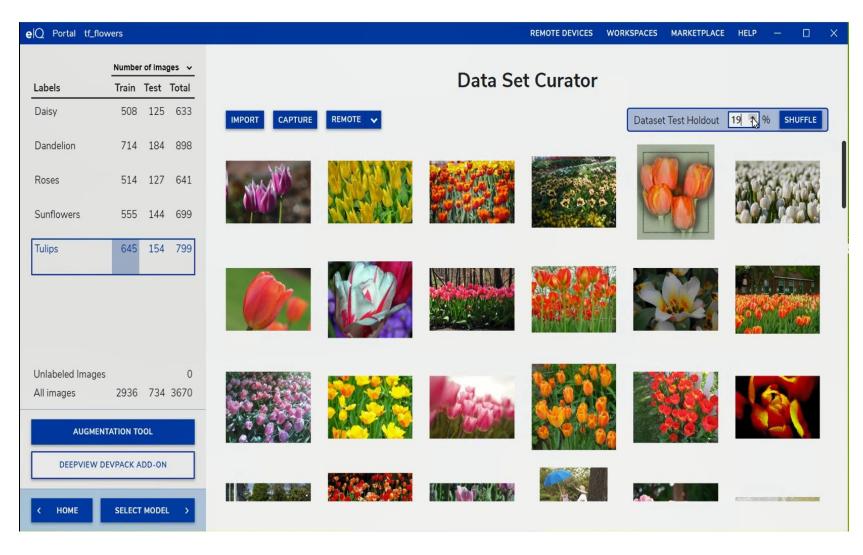
elQ Portal







Data Curation

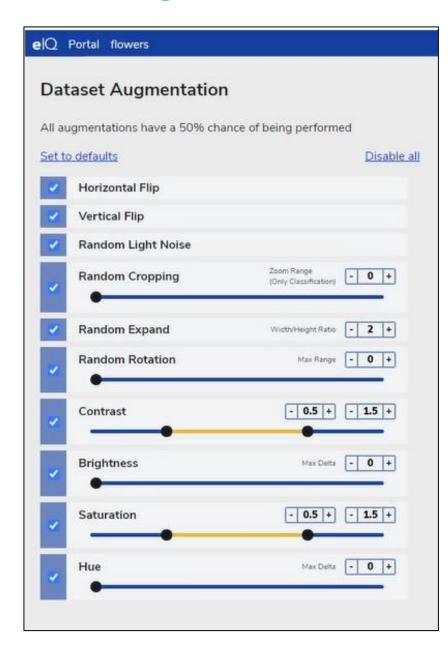


- Capture and annotate images for model training and validation
- Import datasets from public or userdefined formats
- Distribute data for training and testing





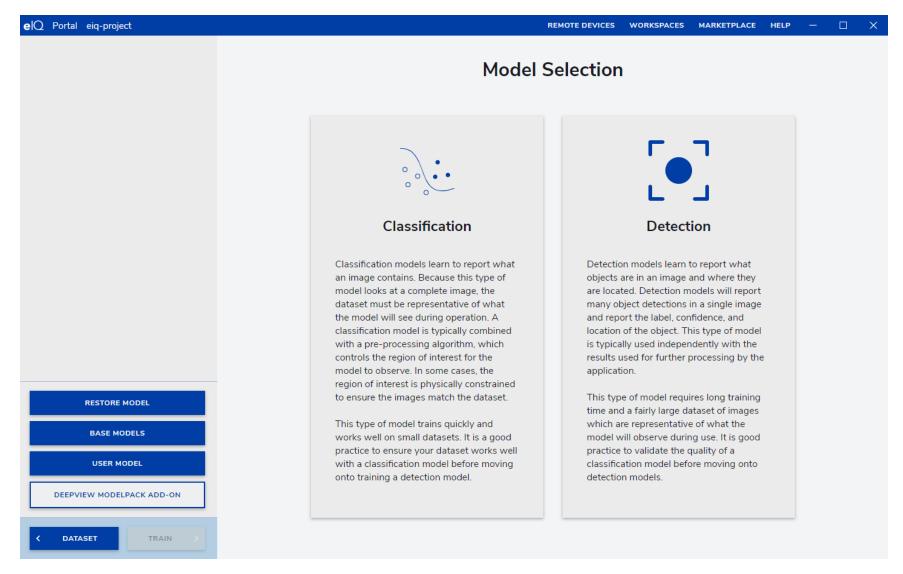
Dataset Augmentation



- Dataset augmentation adjusts image parameters to improve model training
- Reduce over-fitting and increase robustness to dynamic real-world environments
- Visualize how augmentation parameters affect images
- Set augmentation parameters relative to the application



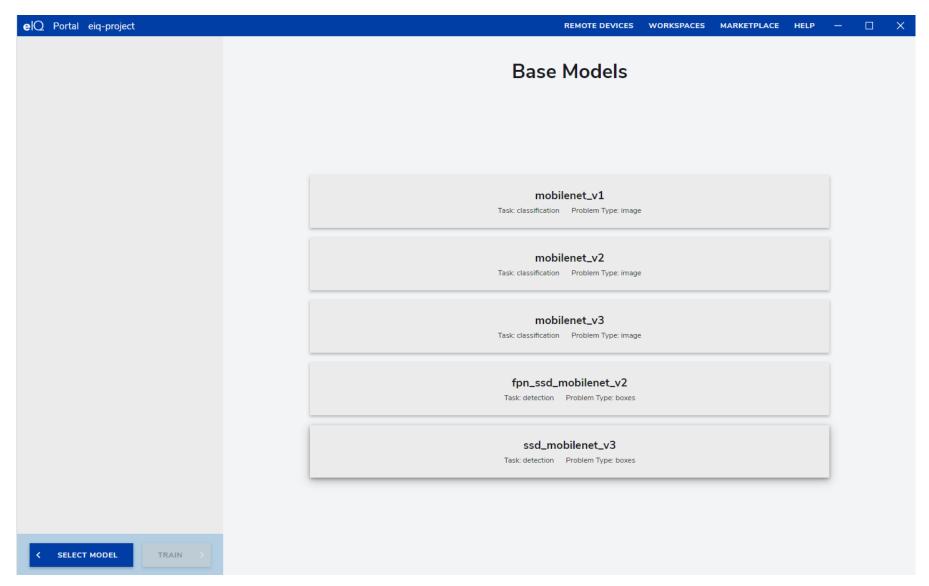
MODEL TOOLS



- Select the appropriate class of model:
 - Classification, Detection

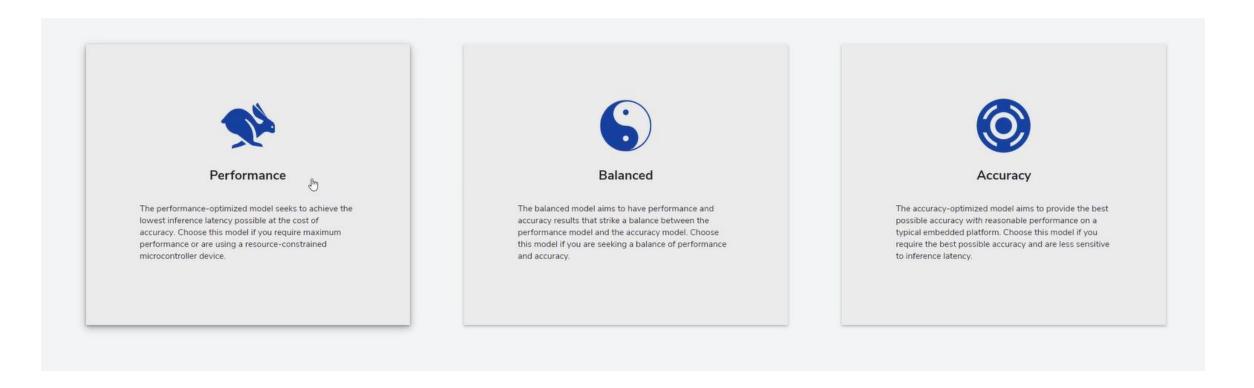


MODEL TOOLS





MODEL TOOLS



 Choose a model version that matches performance and accuracy requirements





MODEL Tools



- Model selected must be correct fit for target resources
 - Application Processors or MCUs
 - Models further optimized for compute units* (MCU, CPU, GPU, NPU)



^{*} Planned future functionality

Model Training and Optimizing for BYOD and BYOM



- Model training should support users of all levels of experience
- Use default settings or fine-tune with hyperparameter selection

- Static model analysis for debug and bottleneck detection
- Automatic graph-level optimizations should improve performance and memory utilization without precision loss (e.g., pruning, fusing, layer folding)
- Configurable optimizations control accuracy tradeoffs (e.g., quantization, layer replacement, weight rounding)
- Quantization converts 32-bit floating point models to 8-bit integer format

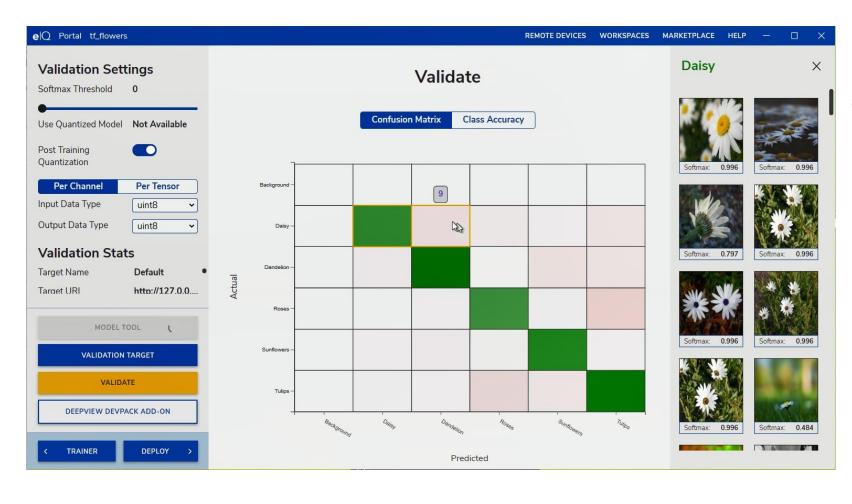








Validation to Prove and Improve Model Behavior

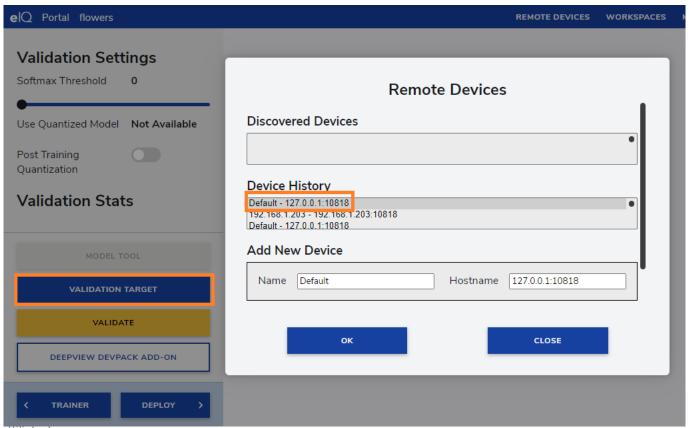


- Use validation to uncover areas that can be improved in your data set
 - Analyze and compare model accuracy running different optimizations



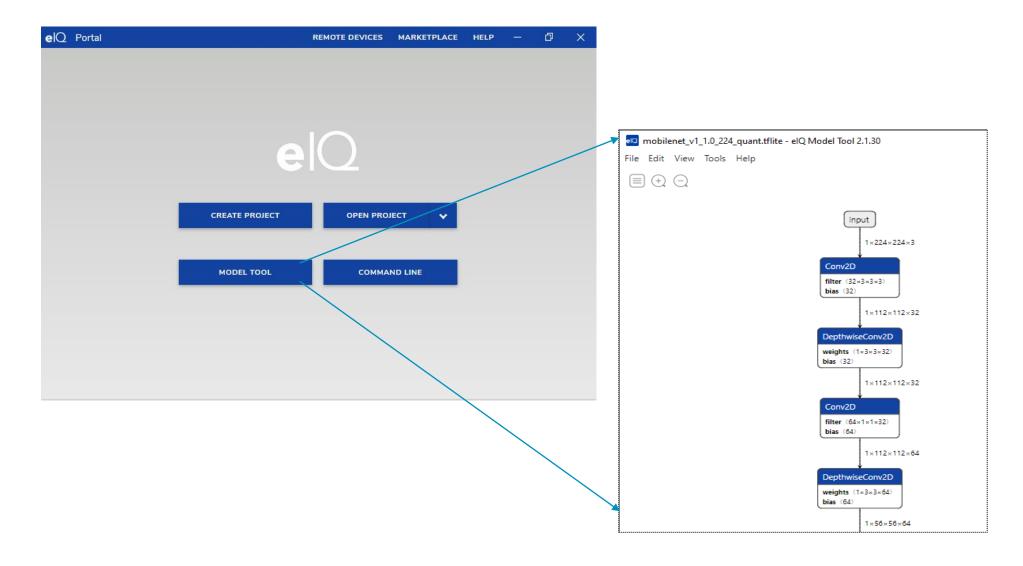
Validation target

- By default you will validate the model on your local PC
- If you have a Remote Device connected you can validate on target and get profiling information





BYOM: eIQ Model tool



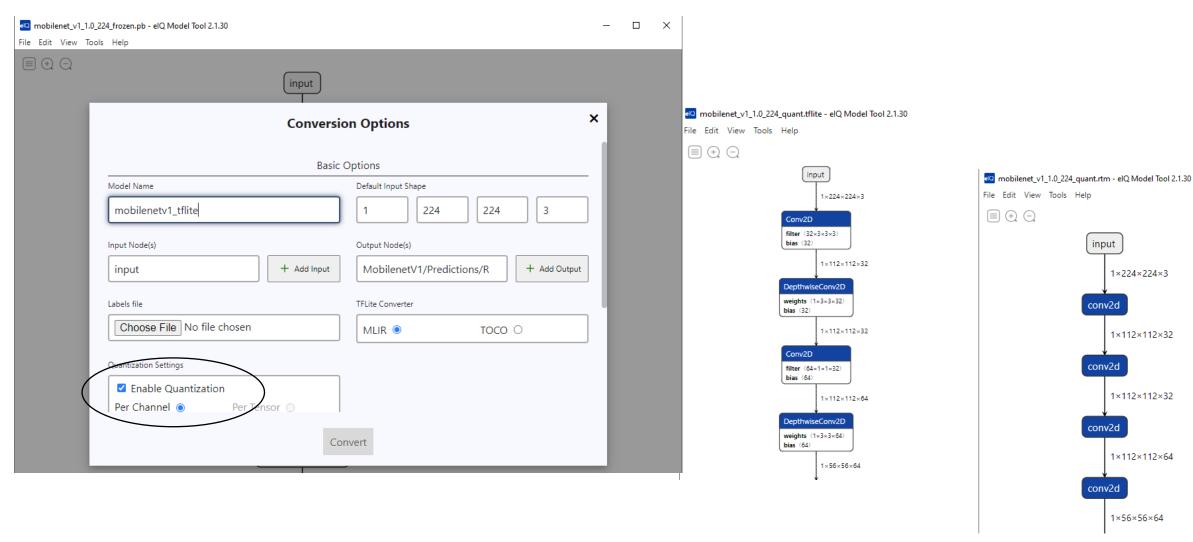


Model converter support

	Float			Quantized		
Source/Destination	DeepviewRT RTM	TensorFlow Lite	ONNX	DeepviewRT RTM	TensorFlow Lite	ONNX
TensorFlow 1.x pb	Yes	Yes	Yes	No	Yes	Yes
Saved Model (Folder/tar)	Yes	Yes	Yes	Yes	Yes	Yes
Keras (.h5)	Yes	Yes	Yes	Yes	Yes	Yes
DeepviewRT RTM	-	No	No	-	No	No
TensorFlow Lite (tflite)	Yes	-	Yes	No	No	Yes
ONNX	Yes	Yes	-	No	Yes	Yes
TensorFlow Lite Quantized	Yes	-	Yes	No	-	-
ONNX Quantized	Yes	Yes	-	No	-	-



elQ Model Converter





1×224×224×3

1×112×112×32

1×112×112×32

1×112×112×64

1×56×56×64

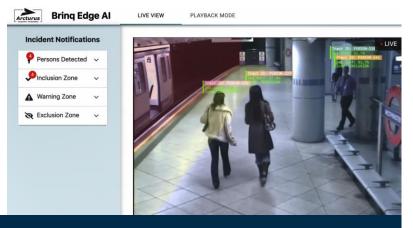


Building and Scaling Applications

David Steele Arcturus

Application Examples



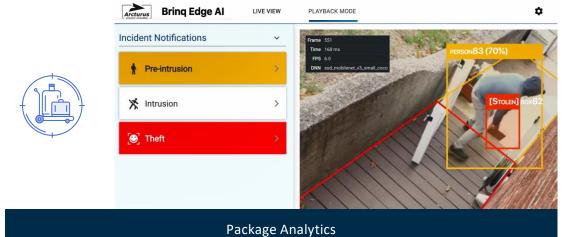


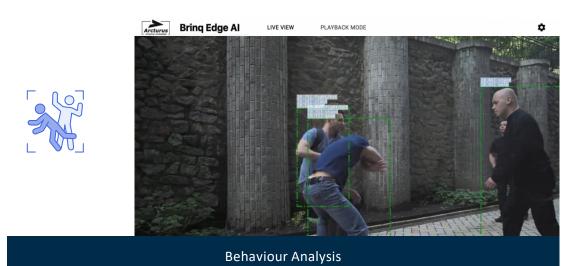
Boundary Crossing and Intrusion





Characterization









Motion Tracking

Motion Tracking Object Detection Classification Model Object Detection Classification Model Object Detection Classification Model Compare / Refine

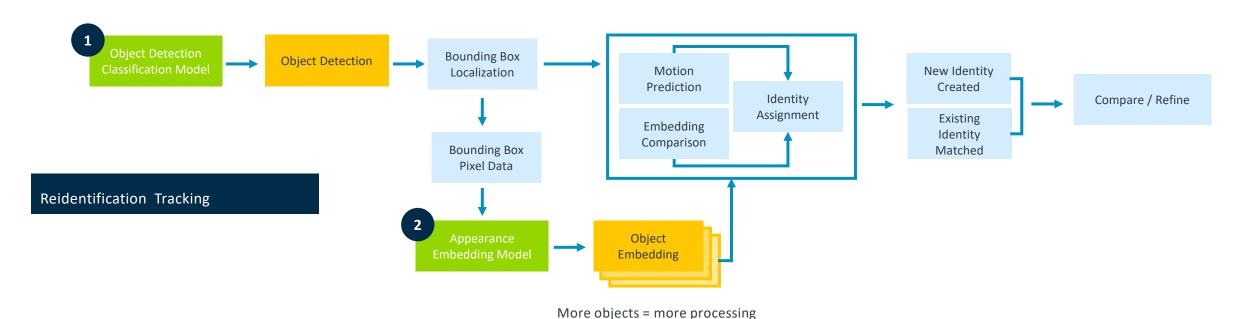
Relies on continuous detections - object cannot leave frame/FoV.





Motion and Appearance Tracking

Motion Tracking



Visual appearance reidentifies objects irrespective of time or space -but- requires generating embeddings for each object detected.







Building and Scaling Pipelines

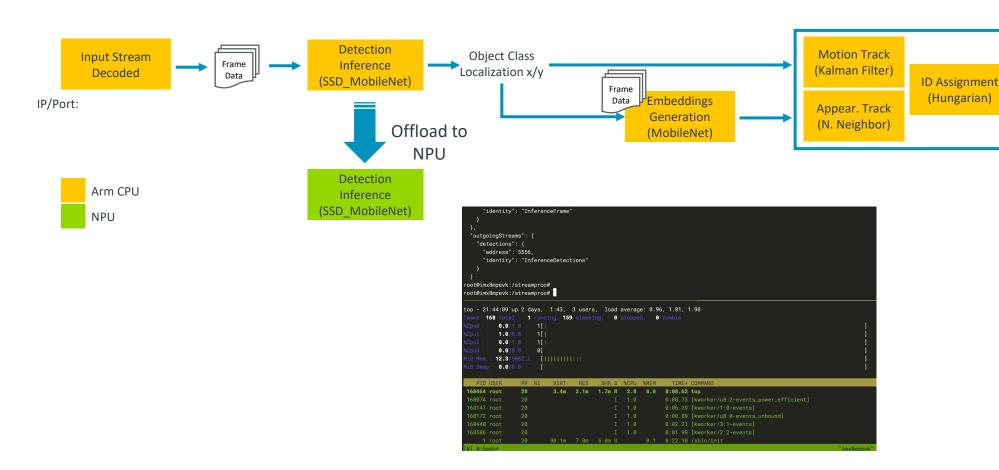
Harshad Mahadik

Edge AI and Vision Team Lead – Arcturus

Jonathan Rynne

Data Scientist – Arcturus

Tracking Pipeline (by resource)









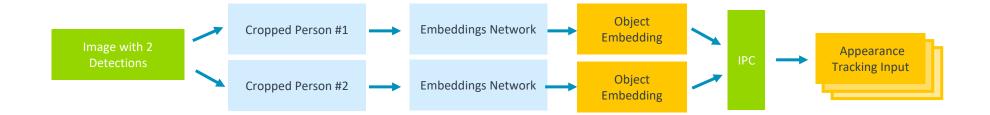
Optimization Techniques

Technique	Description	Benefit	Cost	
Concurrent Processing	Allows multiple models to run at the same time	Decreases latency	Resource intensive	
Batch Processing	Allow multiple images to processed at the same time	Increases throughput	Increases latency	
Model Depth Reduction	Reduces the complexity of the network	Decreases inference time	Decreases accuracy	
Tiling	Breaks down input image into multiple image tiles	Improves accuracy	Increases latency	
Network Quantization	Reduces precision of network e.g. from Float32 to INT8	Improve performance	Decreases accuracy	





Concurrent Model Processing



Pros

- Reduced latency
- Increased throughput
 - Increased FPS
- Increased analytic results

Cons

- Increased complexity
 - memory management (IPC)
- Increased memory consumption
- CPU/GPU/NPU intensive

Performance Example

Mobilenet_v1_embeddings network

- 30ms inference
- 4 People Detected within Frame

Total Inference Time without Concurrent Processing

• 30ms * 4 = 120ms

Total Inference Time with Concurrent Processing

• 30ms + 30ms (overhead) = 60ms

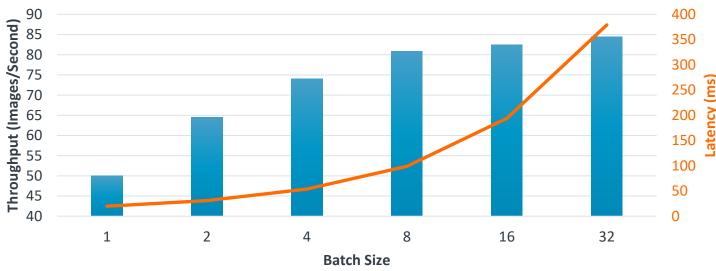




Batch Processing/Batch Inference

Throughput* vs. Latency for Increased Batch Size (PreProcess)





Pros

- Increased throughput
- Decreased complexity

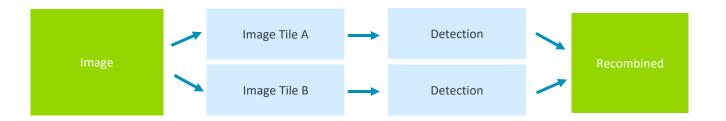
Cons

- Increased memory consumption
- Model/backend support
- Increased latency



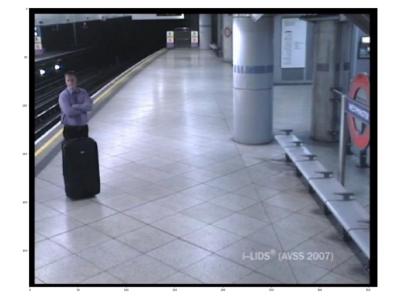


Tiling



Pros

- Improve detection accuracy
- Improve Tracking
- Can be combined with batch



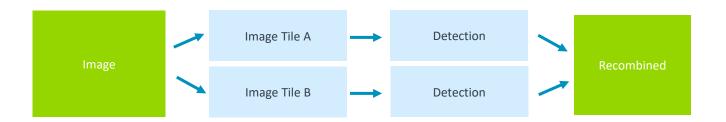






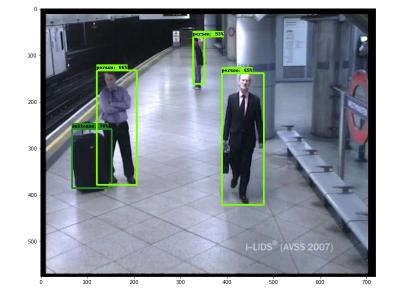


Tiling

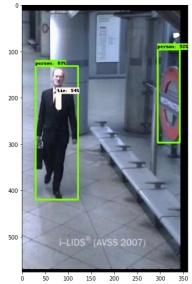


Cons

- Increase false positives
- Add pre/post processing overhead











Model Size Depth Reduction

Definition in <u>source code</u>:

depth_multiplier: Float multiplier for the depth (number of channels) for all convolution ops. The value must be greater than zero. Typical usage will be to set this value in (0, 1) to reduce the number of parameters or computation cost of the model.

Network	Depth Multiplier	Speed* (Embed Time)	Accuracy (mAP on Market 1501)
Mobilenet v1	1.0 (100%)	51ms	63.49%
Mobilenet v1	0.75 (75%)	33ms	62.12%
Mobilenet v1	0.5 (25%)	19ms	60.21%

^{*} Calculated by running model on the CPU (4x Arm Cortex-A53)





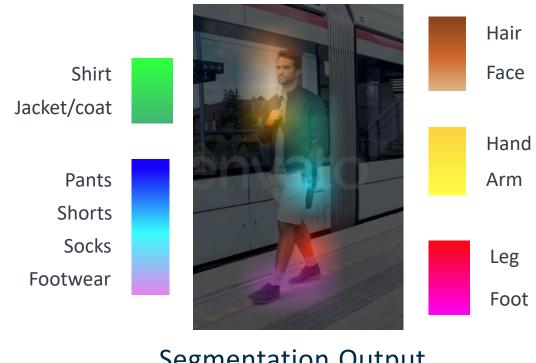
Let's do even more!

David Steele Arcturus

Characterization



Detection and Colour Modeling

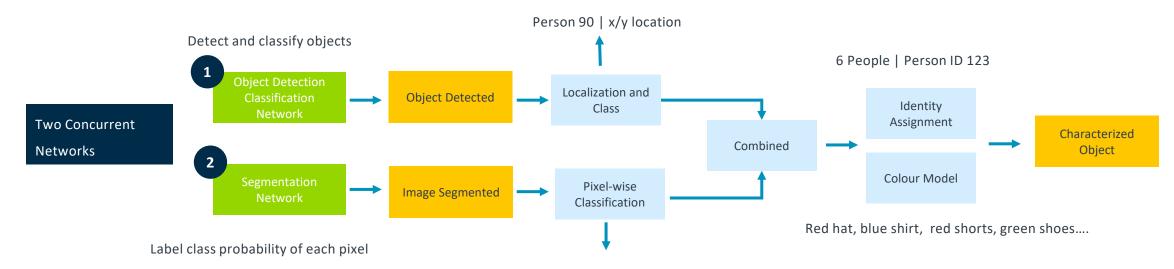


Segmentation Output





Characterization



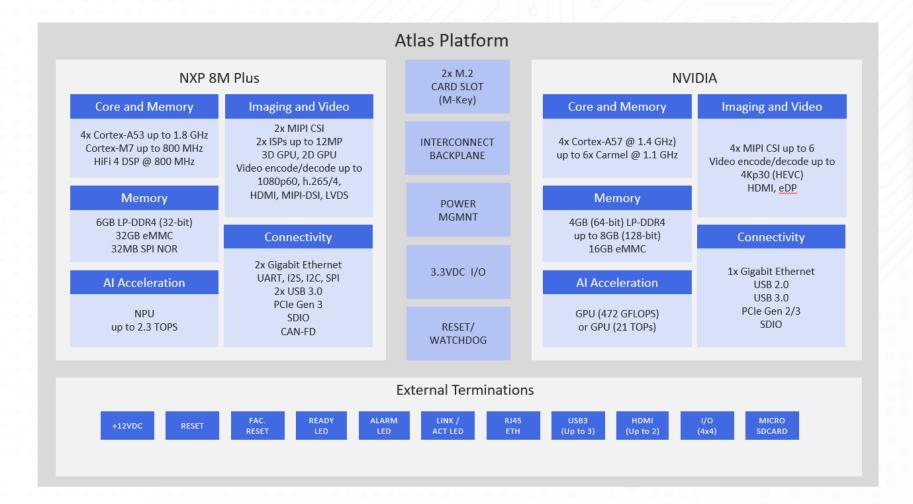
90% Hat, 65% shirt, 40% shorts, 26% pants 5% bikini....

Two networks process the same frame in order to improve precision and fully characterize object





Atlas Edge Al Hardware Platform







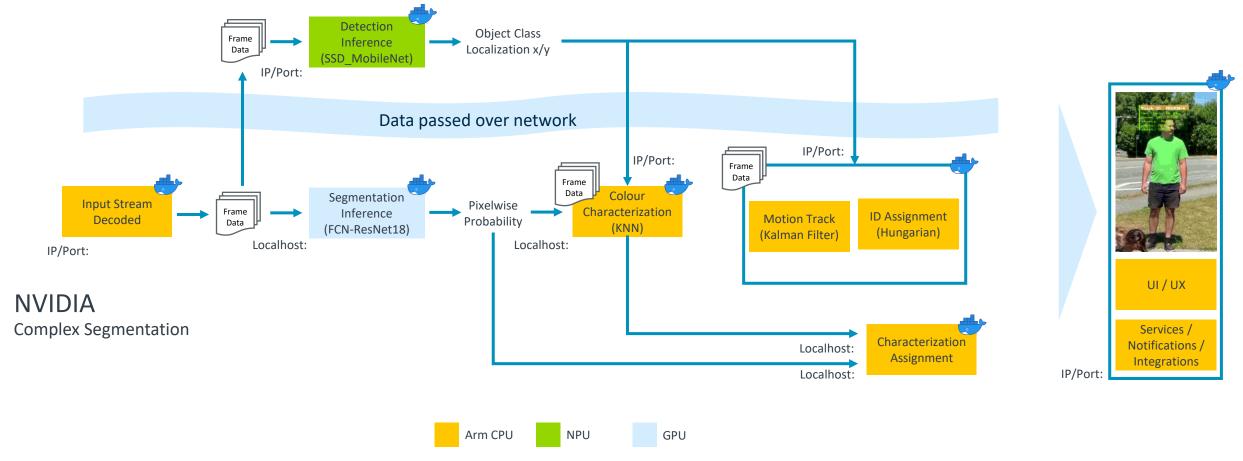




Characterization Pipeline (by resource)

i.MX 8M Plus

High Frame Rate Detection



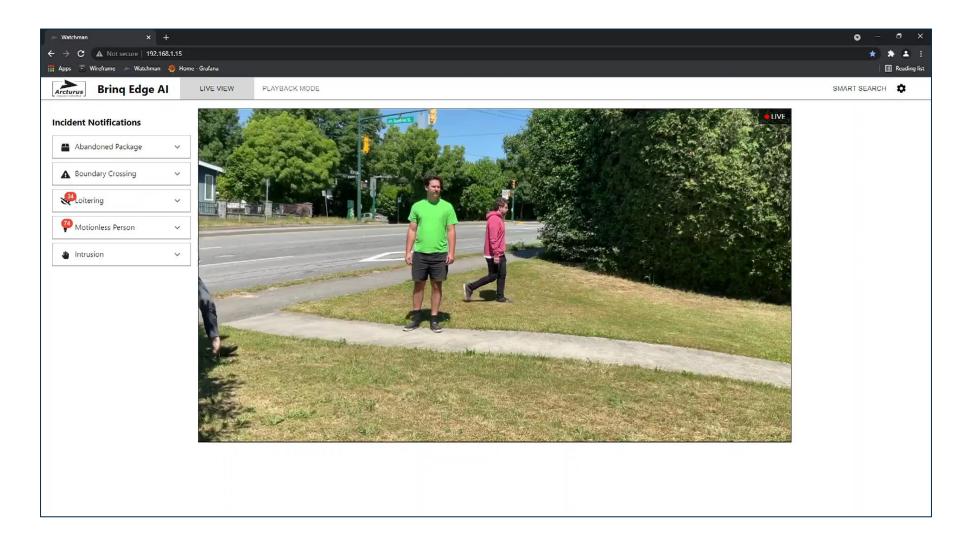






Characterization Demo

Characterization Demo



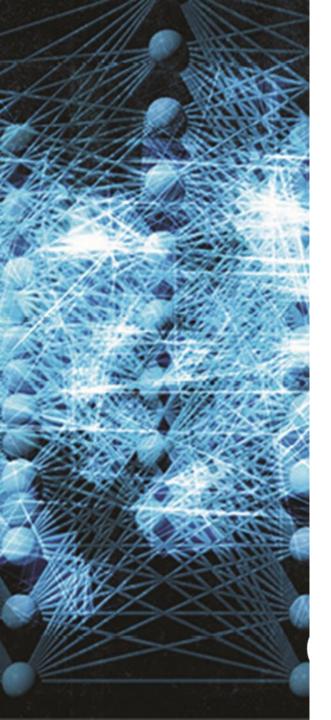




NXP ML/Al training series

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MACHINE LEARNING 20+ training modules Concepts and Introduction **CONCEPTS** Available at www.nxp.com/mltraining eIQ Overview **eIQ™ SOFTWARE DEVELOPMENT** Transfer Learning Intro & Lab **ENVIRONMENT** Handwritten Digit Recognition Example eIQ Toolkit: How to BYOD eIQ TOOLKIT eIQ Toolkit: How to BYOM eIQ Toolkit: Command Line interface 3 eIQ inference with TensorFlow Lite for MCUs - Overview & Lab Machine Learning with i.MX RT MACHINE LEARNING ON MCUS Get started with eIQ on i.MX RT elQ inference with Glow NN Compiler - Overview & Lab Create your own Model using Glow eIQ inference with DeepViewRT eIQ inference with ONNX TensorFlow Lite support for Android ML eIQ inference with Arm NN MACHINE LEARNING ON MPUS eIQ inference with TensorFlow Lite (for MPUs) 5 Solutions and topics contributed PARTNER ML SOLUTIONS by NXP eIQ partners 6



References and helpful links

 eIQ™ ML Software Development Environment (https://www.nxp.com/eiq)

elQ ML/Al Training Series

(https://www.nxp.com/mltraining)

eIQ Community

(https://community.nxp.com/t5/eIQ-Machine-Learning-Software/bd-p/eiq)

Embedded Linux for i.MX Applications Processors

(https://www.nxp.com/design/software/embedded-software/i-mx-software/embedded-linux-for-i-mx-applications-processors:IMXLINUX)

MCUXpresso Software and Tools

(https://www.nxp.com/design/software/development-software/mcuxpresso-software-and-tools-:MCUXPRESSO)

Brinq™ Edge AI and Vision Analytics

(https://www.arcturusnetworks.com/bring/)

Brinq Edge AI for Public Safety

(White paper)

Arm AI Tech Talk – The Smart City In Motion – Intelligent Transportation Systems

(webinar recording)

Arm Dev Summit – Using Arm NN to Develop Edge Alin the Smart City

(webinar recording)







Closing Remarks and Prize Draw



Thank you for participating in our Arm Al Tech Talk.

Complete your information to be eligible to win one of 2x \$250 Amazon Gift cards courtesy of Arcturus and NXP.

The entry deadline is July 13 2021 at 2pm ET. Prize draw will occur on July 13th and winners will be notified by email. One entry per person, per email.

Terms and conditions - by entering you are agreeing to share your information with Arcturus Networks Inc. and NXP Semiconductors for the purpose of contacting you about this and future promotions. Refer to each company's privacy policy and terms of use for additional detail.

* Required







https://forms.gle/ciBtRgC3c76BCNVd7









Thank you!

Tweet us: <u>@ArmSoftwareDev</u>

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Signup now for our next AI Virtual Tech Talk: developer.arm.com/techtalks

Attendees: don't forget to fill out the survey to be in with a chance of winning an Arduino Nano 33 BLE board

Al Virtual Tech Talks Series

Date	Title	Host
July 13 th	Bringing Edge AI to Life - from PoC to Production	Arcturus & NXP
July 20 th	Easy TinyML with Arduino: taking advantage of machine learning right where things are happening	Arduino



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Thank You Danke Merci 谢谢

> ありがとう Gracias

Kiitos 감사합니다 धन्यवाद

شکرًا

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