

The logo for the 2021 Embedded Vision Summit Virtual. It features the year '2021' in a light blue font at the top. Below it, the word 'embedded' is in a smaller, dark blue font. The word 'VISION' is in a large, bold, dark blue font, with the letter 'O' replaced by a colorful circular graphic composed of many small dots. Below 'VISION' is the word 'summit' in a dark blue font. At the bottom, the word 'VIRTUAL' is in a green font, followed by a vertical bar and the dates 'MAY 25-28' in a light blue font. The entire logo is set against a white background with a subtle grid pattern, which is itself centered within a larger graphic of overlapping green and yellow geometric shapes.

2021
embedded
VISION
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VIRTUAL | MAY 25-28

Productizing Edge AI Across Applications and Verticals: NEC Case Study and Insights

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HAILO **NEC**

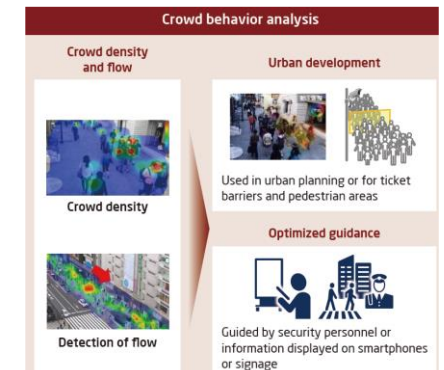
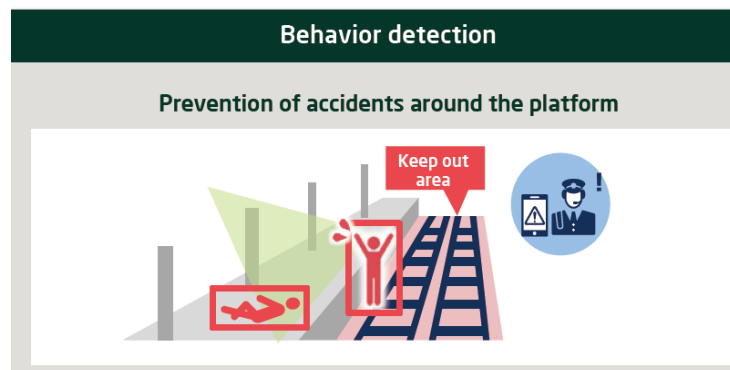
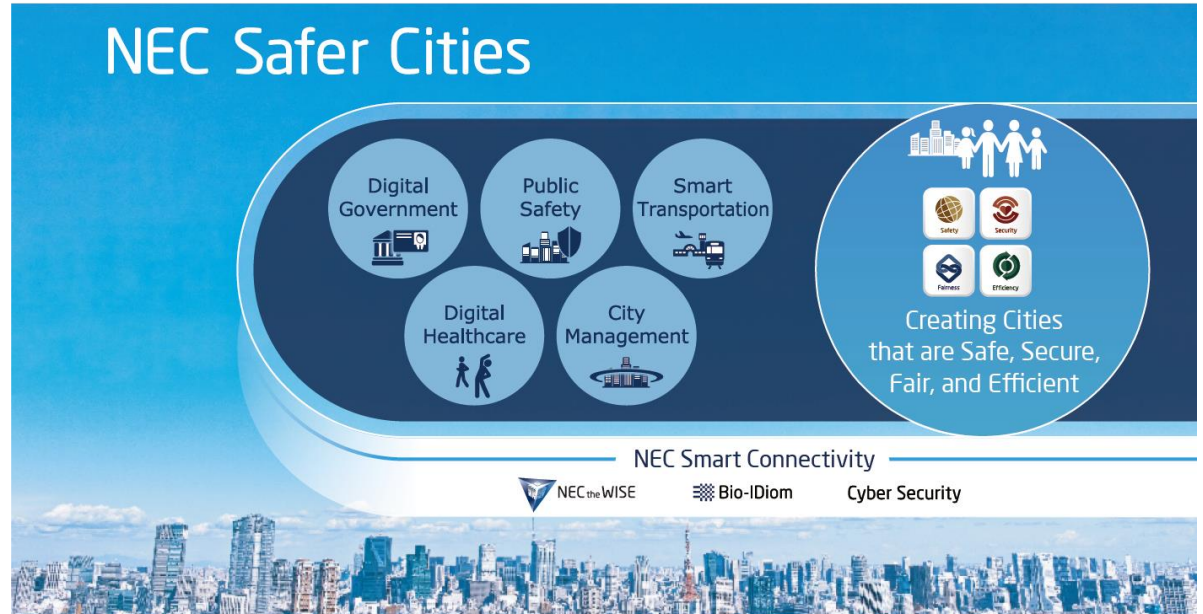
Background

- Started 1 year after company inception
- Initial focus on Public Safety
- Expanded to multiple projects and teams

Key Factors for Success

- Transparency
- Strong and open-minded technical teams
- Clear value to customer

- NEC operates on 5 continents, providing physical safety, failsafe communications and operations solutions
 - Biometrics and video analytics – a major product line
 - Video-based traffic management – a growth area
 - Cost per pixel (camera) has fallen >100X, transmission and storage costs remain high
- Edge video processing is required



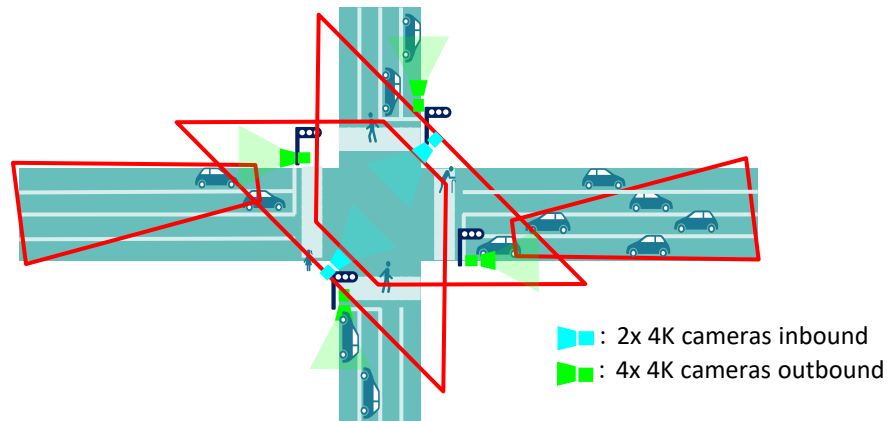
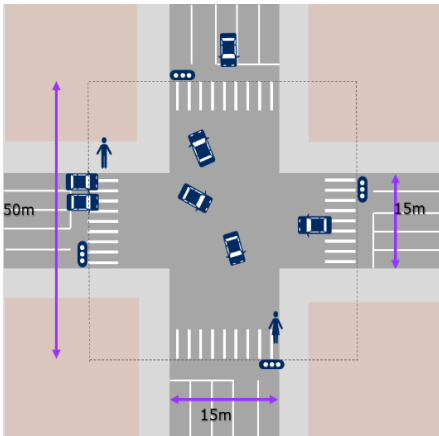
- AI models keep evolving, customers expect “human equivalent” performance
- Making SOTA models run on the edge is not trivial, and most edge AI chips support only a some of the layers, architectures and fixed-point ranges
- Key considerations Include:
 - Can it support the DNN model?
 - Fast CPU interface (bus and drivers)
 - Tight, real-time support for missing features/bugs

Takeaway from Implemented Projects

- Biometrics – the switch to fixed point requires careful QA and precise layer implementation – customers are intolerant to ‘new’ mistakes
- Video Analytics (traffic) – TOPS/W is nice but modern models need a lot of memory for interim layers – sometimes you need to split and context switch
- Video Analytics (safety) – pixels outpace compute – with more compute, even an existing model can work on higher resolution inputs and yield better results. Raw ‘muscle power’ can deliver the better overall performance!

Project Example

- Traffic video analytics: reliable real time recognition of vehicles, pedestrians
- 4 to 6 cameras per junction, in the future: FHD up to 4K
- SOTA YOLO model for object recognition with additional tracking and pedestrian analytics
- A single TensorPC with 2 Hailo-8 cards can reach 4 FHD streams at 30 FPS each
- 4K stresses CPU, not the Hailo-8 cards – this can be addressed with a stronger CPU



- More pixels win – AI works better in higher resolutions. More AI power → better performance
- Fast CPU drivers and data bus are critical
- Conversion of models to the edge is hard, but changing your model is harder – use edge AI that can run your existing models
- Go big or go home – pick a solution that seems “slightly oversize” in compute power and features

- Two system designs to support a range of projects

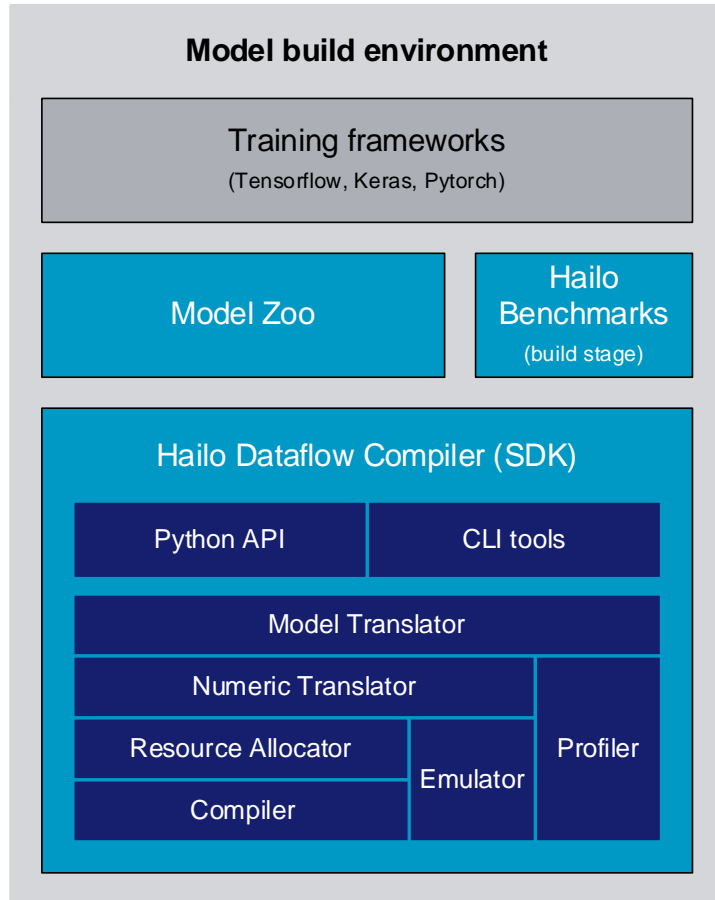
	Mid-Range	High End
Edge Box	Compulab Fitlet2	Compulab TensorPC
Dimensions X*Y*Z [cm]	11.2 * 8.4 * 3.4	20 * 20 * 3.5
Video Interfaces	1/2	4/8/16
AI Performance	26 TOPS (1 module)	26-104 TOPS (1-4 modules)
Power Consumption (typical)	5W-15W	20W-50W



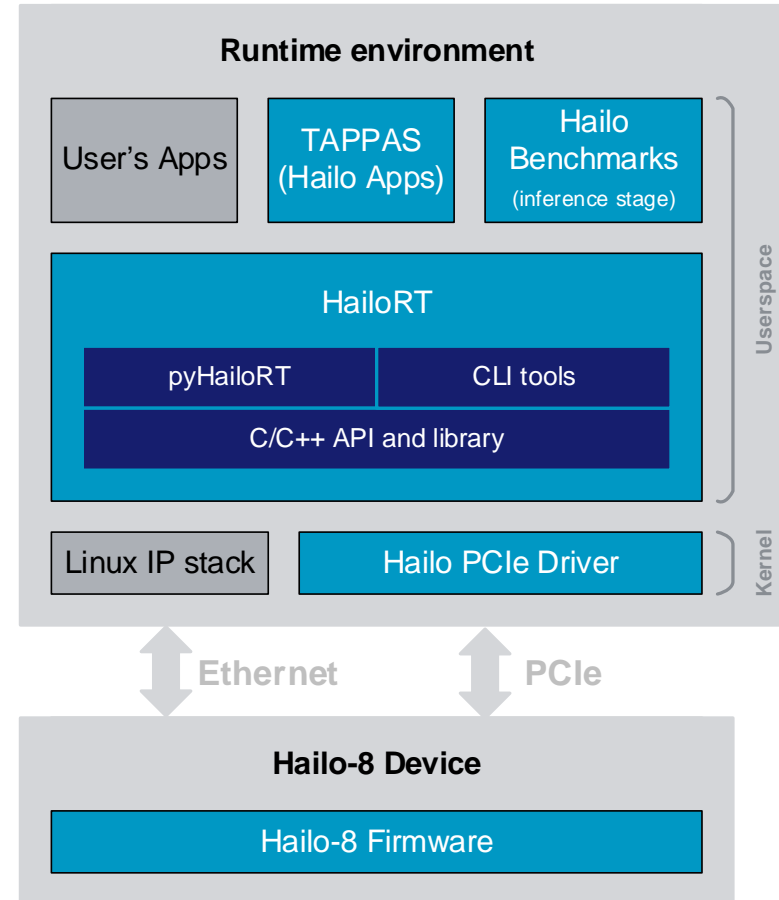
- PCIe interface
- ARM + x86 support
- Power consumption
 - Ex: ResNet-50, 1200 FPS @ 3.8W
 - Near-linear
 - Low power modes support
- Form factors
 - M.2 (A+E, B+M, M)
 - mPCIe (full size)



Build Flow

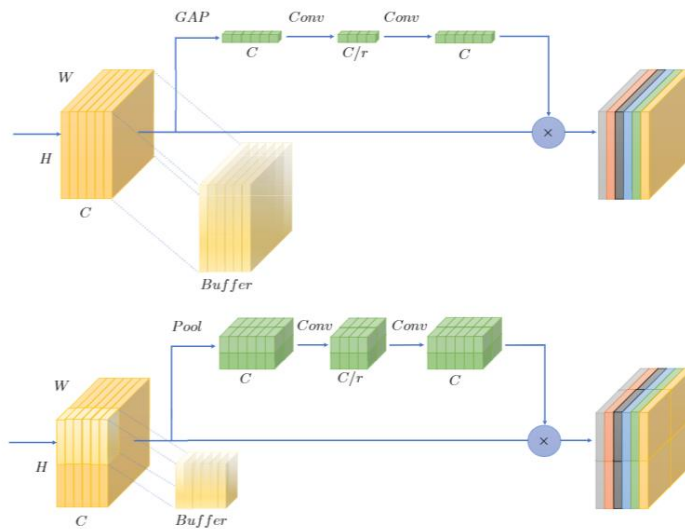


Runtime Flow

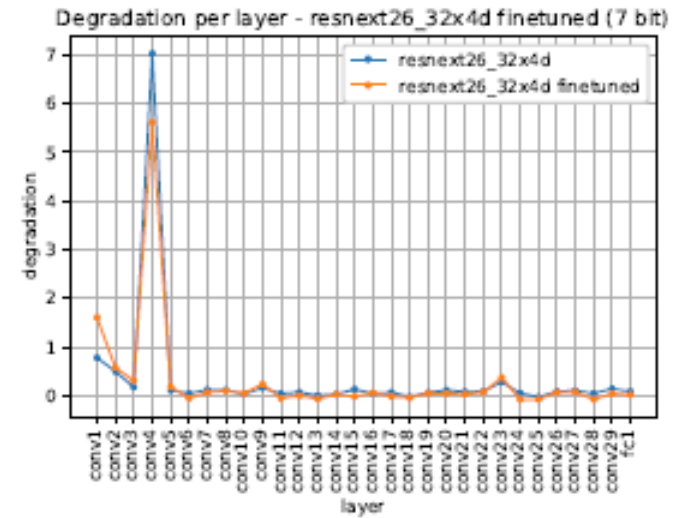


- Hailo software component
- Other software component

- **Early engagement with knowledgeable customer tunes and prioritizes product**
 - Mixed precision and error metrics
 - Squeeze-and-Excite (leading to tiled-SE research)
 - Moving from demos to reference (Hailo [TAPPAS](#))



- **Understanding customer development flow**
 - Provide tools that bring our ML expertise into customer hands ([example](#) – LAT)
- **Customer needs prioritize highly-optimized models**
 - Inputs to Hailo's model zoo roadmap
- **Roadmap refinement (device N+2 effect)**
- **Compute requirements only increase...**



HAILO PRODUCTS INDUSTRIES TECHNOLOGY RESOURCES COMPANY BLOG

What's new Halo SW TAPPAS Model Zoo Documentation Benchmarks

Model Zoo

The Hailo model zoo provides deep learning models for various computer vision tasks. The pre-trained models can be used to create fast prototypes on Hailo devices. Main features:

- Variety of pre-trained models and tasks in TensorFlow and ONNX.
- Providing model details including full precision accuracy vs. quantized model accuracy which was measured on Halo-8™.
- Each model also includes a binary HEF file that is fully supported in the Hailo toolchain and Application suite (for registered users only).

Supported Vision Tasks

- Classification >
- Object Detection >
- Segmentation >
- Pose Estimation >
- Head Pose Estimation >
- Driving Lane Detection >
- Multiple Object Tracking >
- Face Recognition >