



A View from the Summit (Part 2)

Jeff Bier
Founder, Edge AI and Vision Alliance
President, BDTI
May 18, 2022

What I Learned Yesterday



Manipulating Audio from Time Domain to Frequency Domain

A lot of techniques employed for ML audio-based solutions borrow techniques from vision. This means we need to take a 1D sequence of samples and make them look like an image.

[https://en.wikipedia.org/wiki/Sampling_\(signal_processing\)](https://en.wikipedia.org/wiki/Sampling_(signal_processing))
<https://upload.wikimedia.org/wikipedia/commons/c/c5/Spectrogram-1984C.png>
© 2022 DSP Concepts

DSP Concepts

Josh Morris, DSP Concepts



What I Learned Yesterday



Manipulating Audio from Time Domain to Frequency Domain

A lot of techniques employed for ML audio-based solutions borrow techniques from vision. This means we need to take a 1D sequence of samples and make them look like an image.

[https://en.wikipedia.org/wiki/Sampling_\(signal_processing\)](https://en.wikipedia.org/wiki/Sampling_(signal_processing))
<https://upload.wikimedia.org/wikipedia/commons/c/c5/Spectrogram-1984C.png>

DSP Concepts © 2022 DSP Concepts

Josh Morris, DSP Concepts

Multi-sensor fusion strategies

- Early fusion**
Fuse sensor data and then perform inference using a network
- Late fusion**
Perform inference from each sensor data and then merge the predictions
- Mid-level fusion**
Fuse intermediate representations from sensor data and then train a predictor
- Sequential fusion**
Use sensor data inference in sequence to refine predictions
See FrustrumNet paper in references

SENSOR CORTEK 2022 University of Ottawa — Sensor Cortek Inc.

Robert Laganieri, Sensor Corek



What I Learned Yesterday



Manipulating Audio from Time Domain to Frequency Domain

A lot of techniques employed for ML audio-based solutions borrow techniques from vision. This means we need to take a 1D sequence of samples and make them look like an image.

[https://en.wikipedia.org/wiki/Sampling_\(signal_processing\)](https://en.wikipedia.org/wiki/Sampling_(signal_processing))
<https://upload.wikimedia.org/wikipedia/commons/c/c5/Spectrogram-1984C.png>

© 2022 DSP Concepts

Josh Morris, DSP Concepts

Multi-sensor fusion strategies

- Early fusion**
Fuse sensor data and then perform inference using a network
- Late fusion**
Perform inference from each sensor data and then merge the predictions
- Mid-level fusion**
Fuse intermediate representations from sensor data and then train a predictor
- Sequential fusion**
Use sensor data inference in sequence to refine predictions
See FrustrumNet paper in references

© 2022 University of Ottawa — Sensor Corek Inc.

Robert Laganieri, Sensor Corek

Training the Teacher

$$H(p, \hat{p}) = - \sum_i p_i \log \hat{p}_i$$

- Cross-entropy loss: $H(p, \hat{p})$
- Ground-truth and predicted probabilities p, \hat{p}

© 2022 Bending Spoons

Federico Perzi, Bending Spoons

What I Learned Yesterday



Manipulating Audio from Time Domain to Frequency Domain

A lot of techniques employed for ML audio-based solutions borrow techniques from vision. This means we need to take a 1D sequence of samples and make them look like an image.

[https://en.wikipedia.org/wiki/Sampling_\(signal_processing\)](https://en.wikipedia.org/wiki/Sampling_(signal_processing))
<https://upload.wikimedia.org/wikipedia/commons/c/c5/Spectrogram-1984C.png>

DSP Concepts © 2022 DSP Concepts

Josh Morris, DSP Concepts

Multi-sensor fusion strategies

- Early fusion**
Fuse sensor data and then perform inference using a network
- Late fusion**
Perform inference from each sensor data and then merge the predictions
- Mid-level fusion**
Fuse intermediate representations from sensor data and then train a predictor
- Sequential fusion**
Use sensor data inference in sequence to refine predictions
See FrustrumNet paper in references

SENSOR CORTEK © 2022 University of Ottawa — Sensor Cortek Inc.

Robert Laganriere, Sensor Corek

Training the Teacher

Classification Dataset

teacher

logits

probabilities

hard labels

CE Loss

$$H(p, \hat{p}) = - \sum_i p_i \log \hat{p}_i$$

- Cross-entropy loss: $H(p, \hat{p})$
- Ground-truth and predicted probabilities p, \hat{p}

BENDING SPOONS © 2022 Bending Spoons

Federicao Perzzi, Bending Spoons

Tissue Foundry: a prototype Scalable, Modular, Automated, Closed manufacturing system

Tissue Harvest and Cell Banking

Expansion Culture

Cell Harvest and Wash

Scaffold Fabrication

Tissue Assembly and Maturation

Preservation and Packaging

Transport and Logistics

Raman spectroscopy (metabolites)

Capacitance (# live cells)

Optical flow cells (pH, Dissolved Oxygen, Glucose)

armi biofabuSA

Dean Kamen





Opportunities: Smart Spaces

Almost Any Space Can Be a Smart Space



Consumer



- Homes
- Apartments
- Condominiums
- Front yards
- Back yards
- Storage

Commercial



- Retail
- Industrial
- Offices
- Agriculture
- Healthcare

Government/Municipal



- Highways, roadways
- Ports, airports
- Recycling, waste processing
- Schools, campuses
- Public transit
- Street and garage parking
- Parks

Smart Space Applications Have Blossomed



Source: Petcube



Source: cpb.gov



Source: Compology



Source: Ringdoorbell.eu



Source: Orbital Insight



Source: Identified Technologies



Source: Amazon



Source: Hortdaily.com

Benefits of Smart Spaces



Smart Spaces deliver three big potential benefits:

User experience (e.g., increases comfort or convenience)

Economics (e.g., boosts efficiency or reduces cost)

Safety/risk (makes a place safer or less risky)

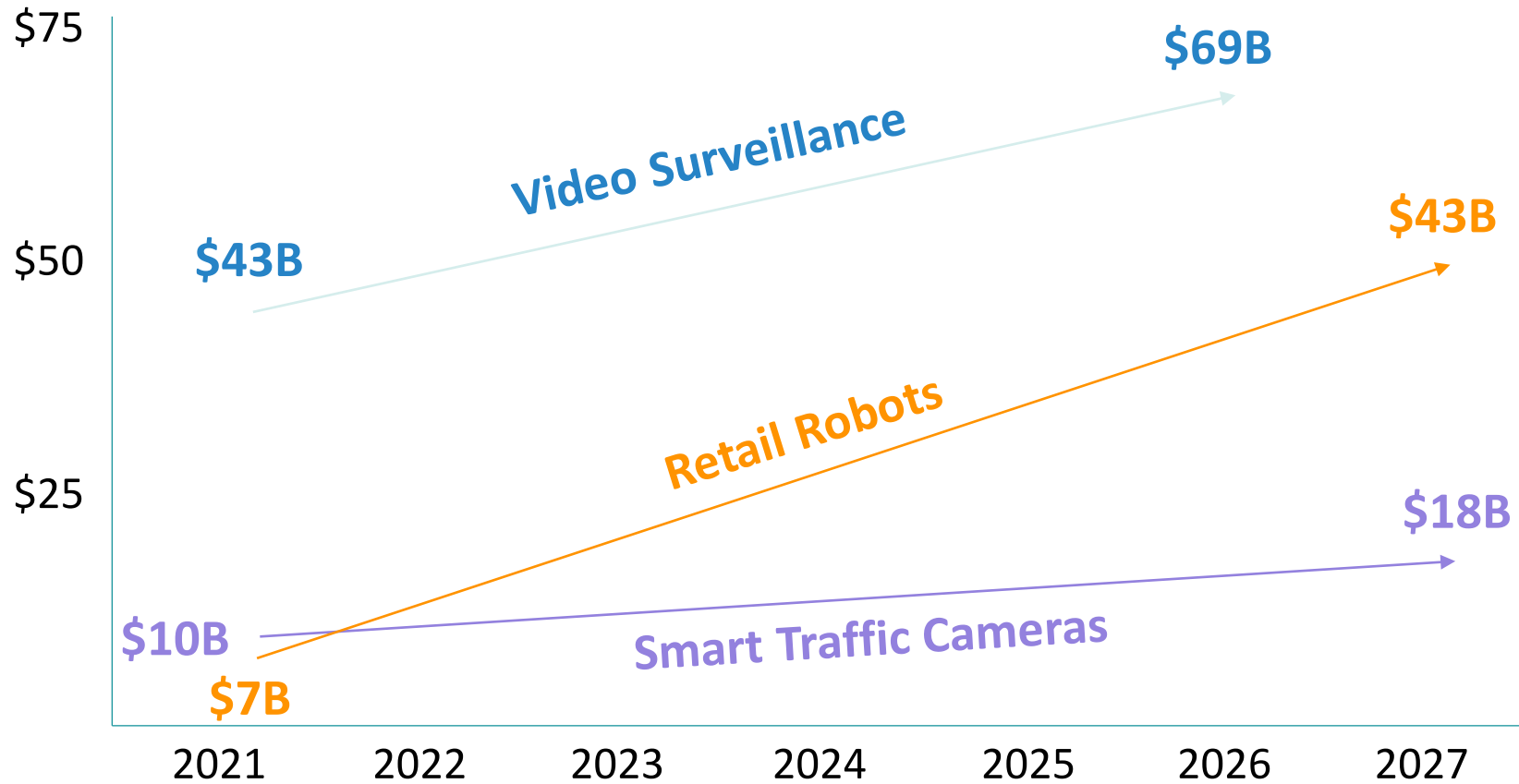
Smart Spaces Success Story: Cubic GRIDSMART



The Smart Spaces Market Is Big ... and Growing



USD billions



Smart Spaces broadly encompasses many markets ... many of which are large in their own right!

Sources: Markets and Markets, KBV Research, Coherent Market Research, Grand View Research; Alliance estimates

- **Sensors**
 - Image/video
 - Audio
 - Optical depth
 - RF (e.g., tracking via phones)
 - Radar
 - Air quality
 - Vibration
- **Processing**
 - Edge processors
 - Cloud computing services
- **Connectivity**
 - Wireless communications
- **Algorithms**
 - Signal conditioning
 - E.g., high dynamic range imaging
 - Perception
 - E.g., audio and visual AI, tracking, intent prediction, ...
 - Privacy preservation
 - E.g., face blurring
 - Sensor fusion
- **Frameworks and tools**
 - DevOps
 - Monitoring, updates
- **Security**

Key Challenges in Smart Spaces



- Lack of expertise
- Training data
- Cost of deployment
- Fragmentation / long tail
- Accuracy in diverse and challenging environments
- Reliability
- Sensor fusion
- Aggregation of data and/or analytics from distributed devices
- Edge-cloud processing trade-offs
- Managing large distributed systems
- Privacy
- Others?

Observation: Privacy



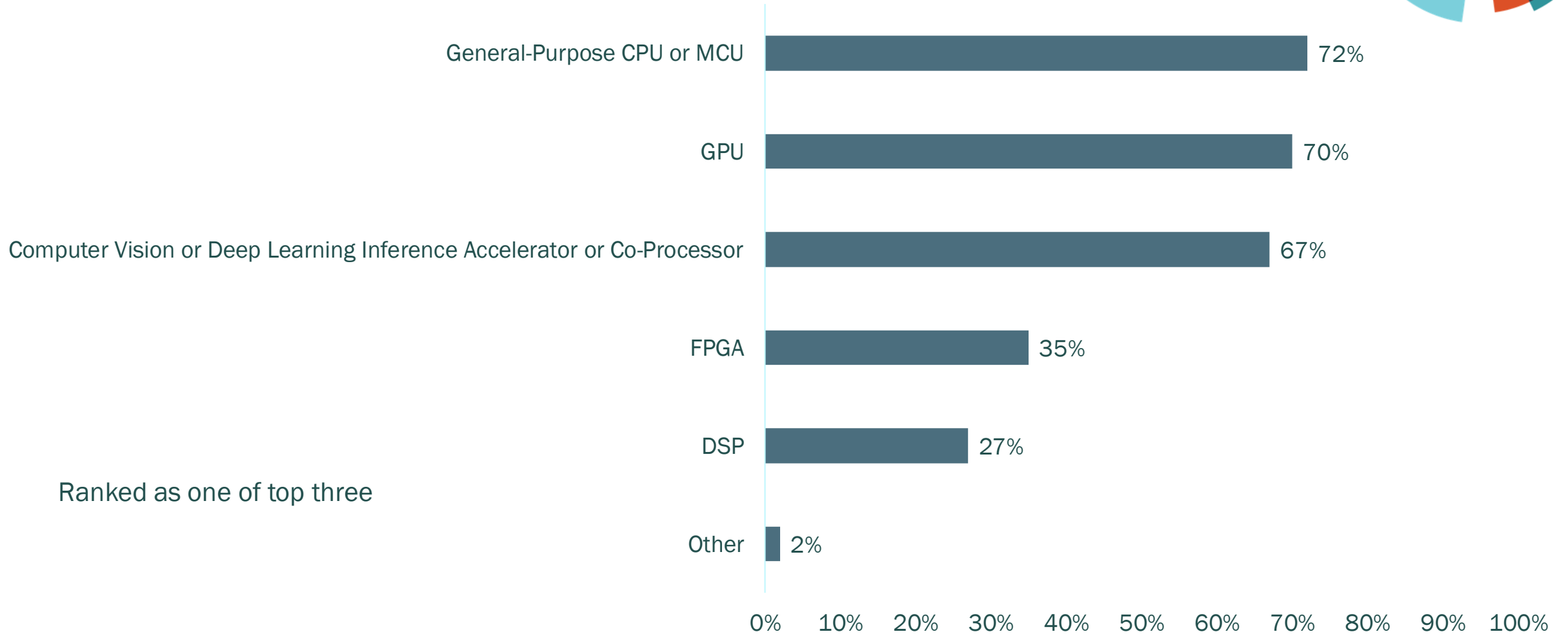
- Smart Spaces offer a potential future of improved user experience, economics, and safety
- They also offer a potential future of ubiquitous surveillance
- Regulation on various aspects of Smart Spaces (such as facial recognition) is here, and growing
- A “privacy first” approach to Smart Spaces will be important

German coalition backs ban on facial recognition in public places

Maine passes the strongest state facial recognition ban yet

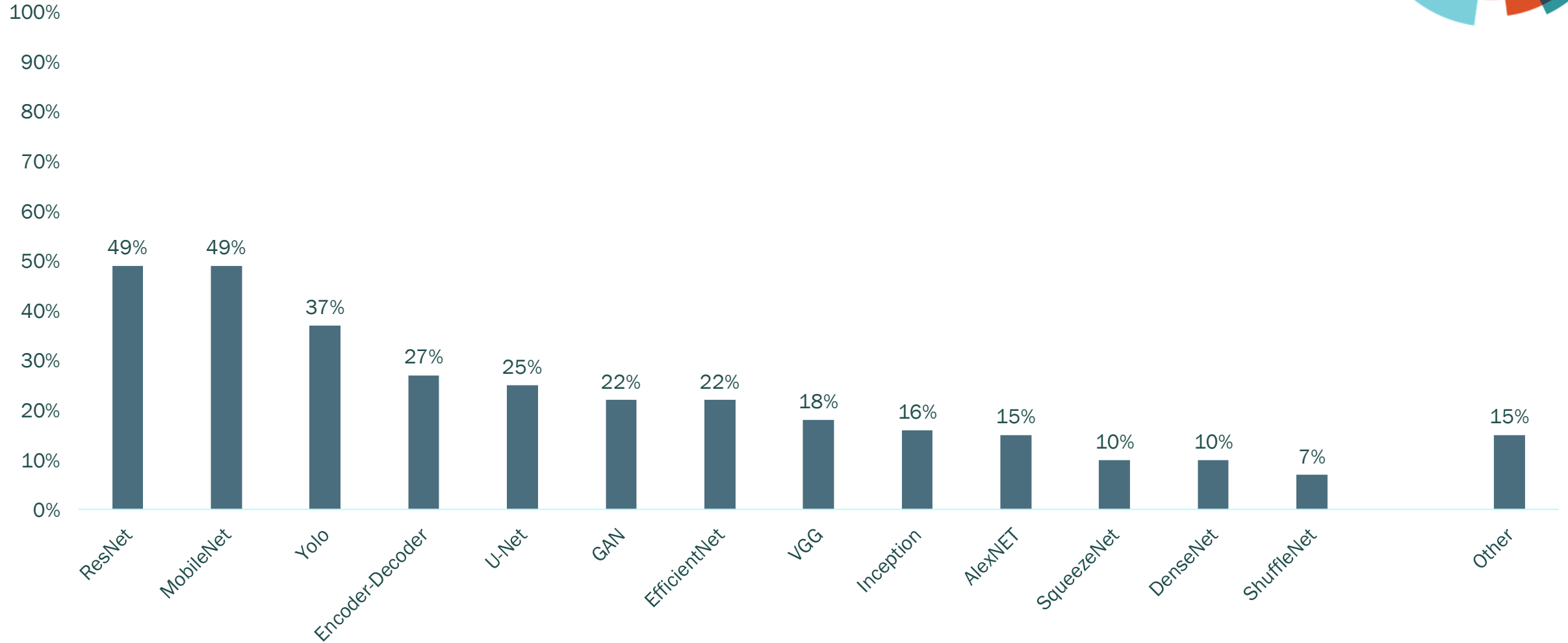
European Parliament backs ban on remote biometric surveillance

Type of Processor Used for Deployment of Vision Tasks



Source: Edge AI and Vision Alliance, *Computer Vision Developer Survey, November 2021*

DNN Topology



Source: Edge AI and Vision Alliance, *Computer Vision Developer Survey, November 2021*



FIRST Robotics Competition

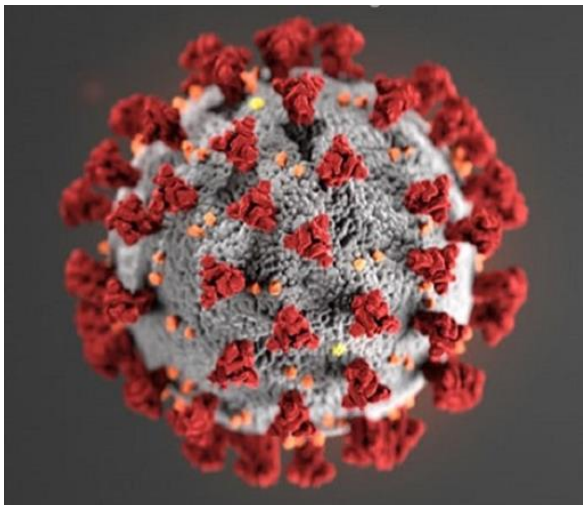
FIRST Robotics Competition Honorees





Making the Most of Your Day

Let's Be Safe



Lohp.Berkeley.edu



USDA

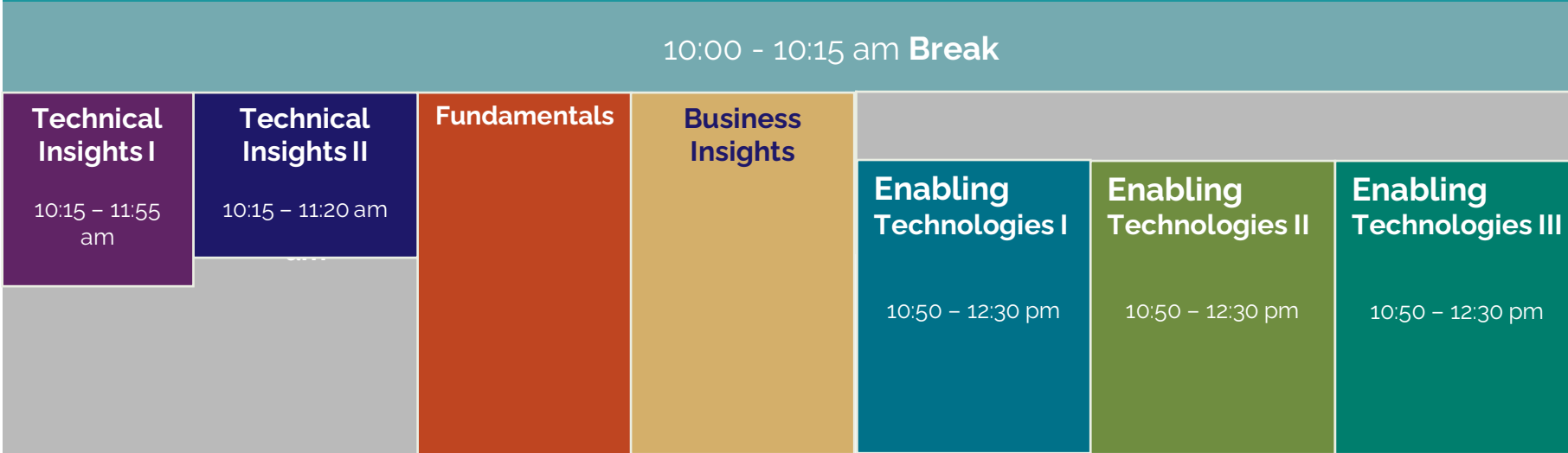
Wednesday Schedule



9:00 - 10:00 am

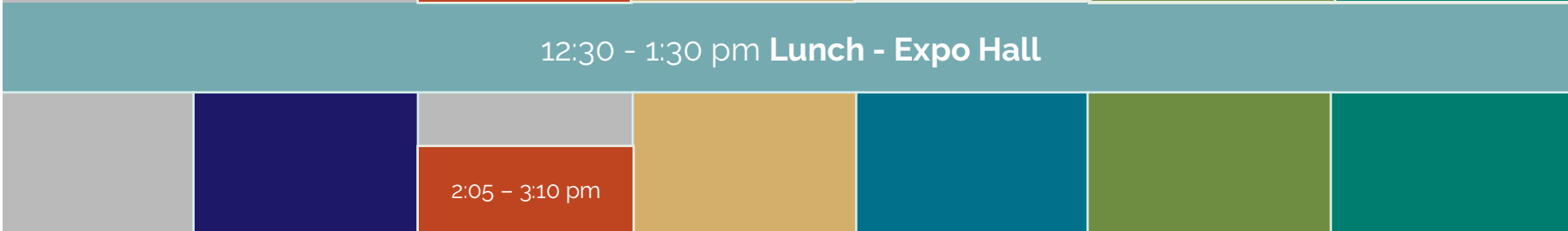
INTRODUCTION: A View from the Summit (Part 2) – Jeff Bier
GENERAL SESSION: Powering the Connected Intelligent Edge and the Future of On-Device AI – Ziad Asghar

10:15 am - 12:30 pm

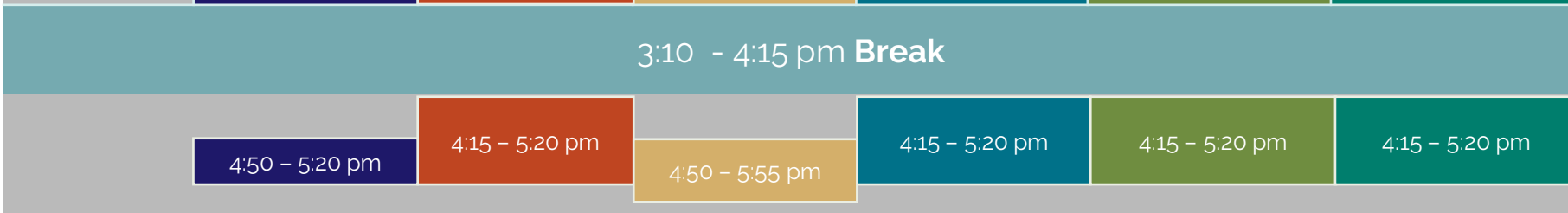


10:00 am - 6:00 pm
Technology Exhibits
 Expo Hall

1:30 - 3:10 pm



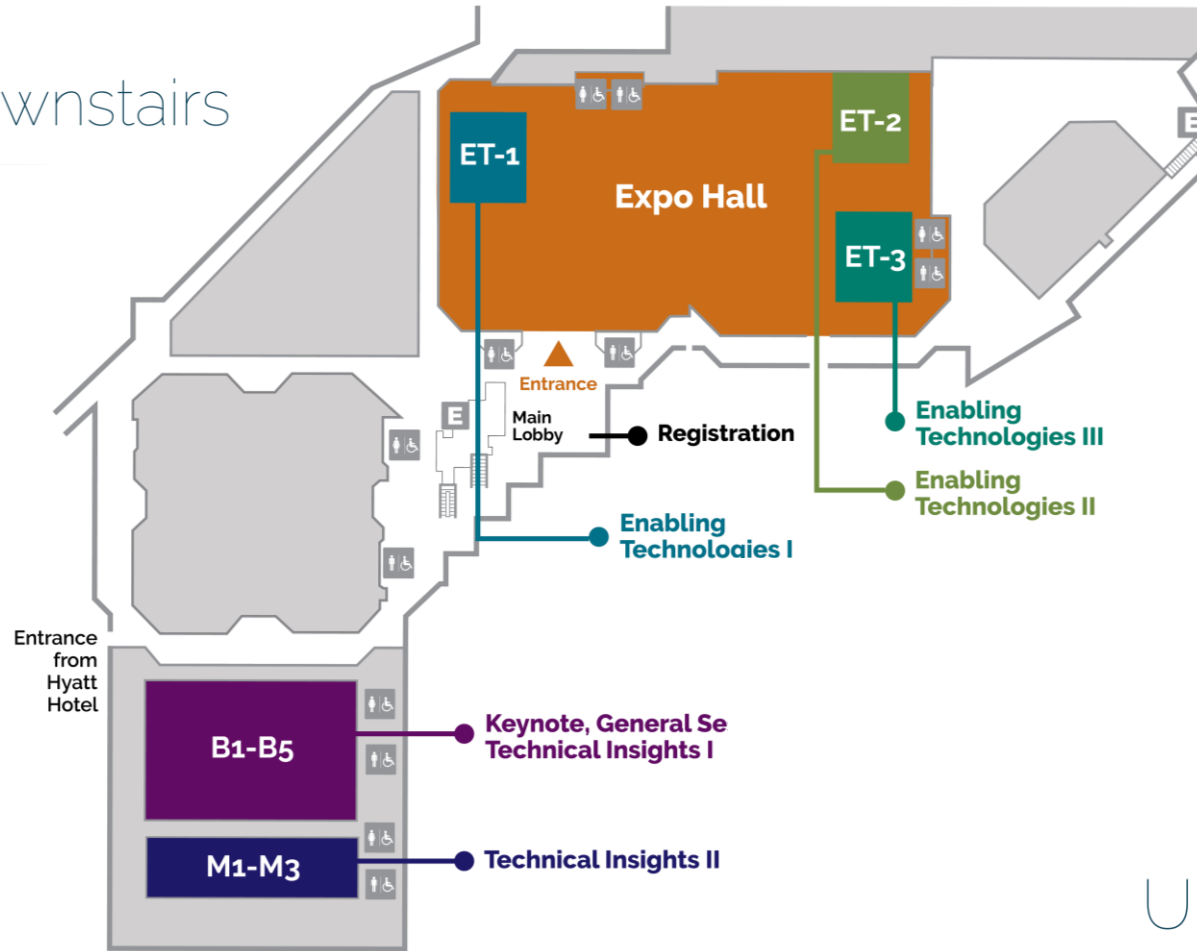
4:15 - 6:00 pm



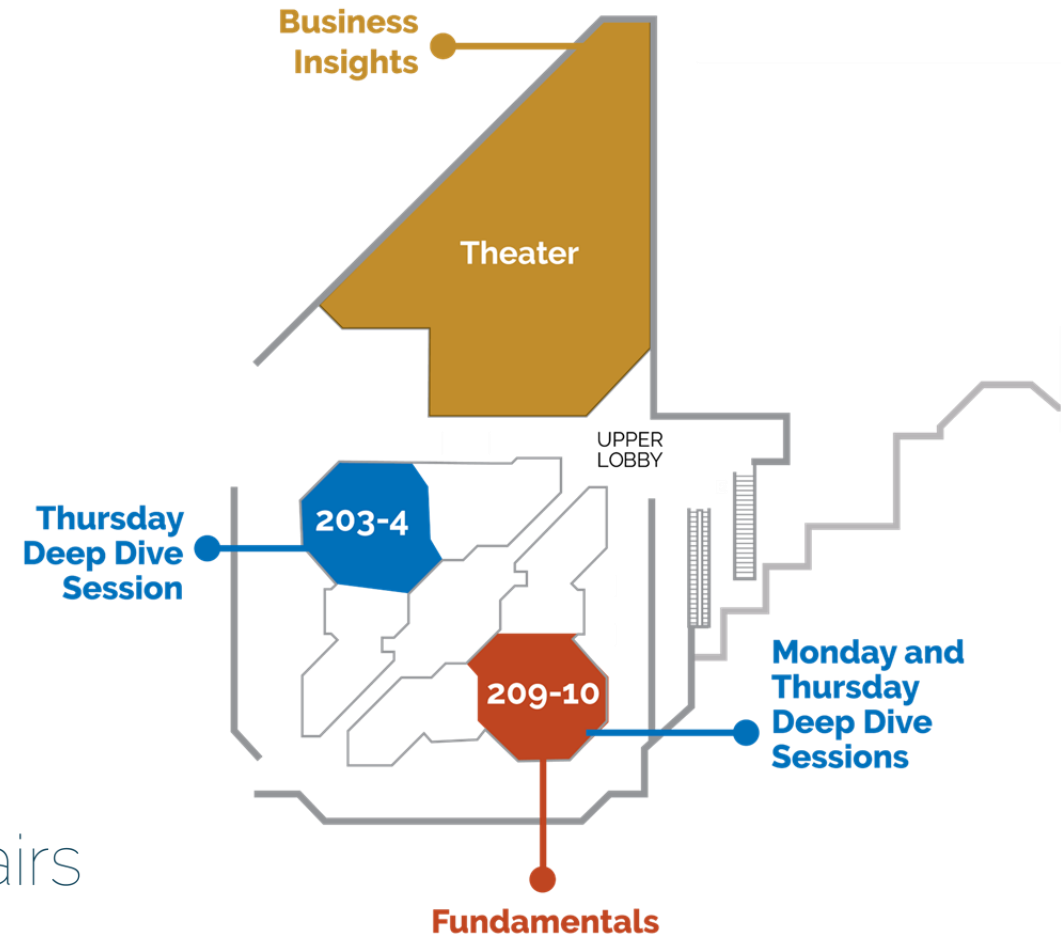
Facility Map



Downstairs



Upstairs



Check Out the Exhibits!



Deep Dive Day Sessions - Thursday



Develop and Deploy Advanced Edge Computer Vision — Fast!

9:00 am - 12:00 pm



Optimize AI Performance and Power for Tomorrow's Neural Network Applications

12:00 - 3:00 pm



Intel AI Developer Expo — Let's Build Something Wonderful Together

Session
3:00 - 5:30 pm

Reception
5:30 - 7:30 pm

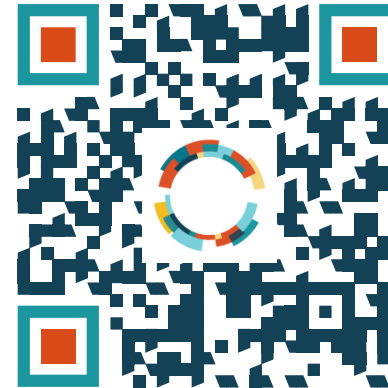
Visit the Registration desk to reserve your seat!



Download Our App and Get the Full Summit Agenda on Your Phone!



Create a personal agenda of events, find demos and exhibitors in the Expo Hall and stay up to date on the latest Summit news through our new mobile app.



It's as easy as

1. Download the app

For either iPhone or Android, scan the QR code with your phone.

QR code not working?

Visit uqr.to/2022summit or search for Embedded Vision Summit in the App Store or the Play Store.

2. Search your email

Look for an email from noreply@mg.gripcontact.com which will have your unique login details

3. If you cannot find the email

Open the app and enter the email address with which you registered for the Summit and the confirmation number you received from registration@edge-ai-vision.com to set up your account





Please fill out your Embedded Vision Summit survey!

Completed surveys will be entered into a drawing for one of three \$100 Amazon gift cards.



Thank You!



Powering the Connected Intelligent Edge and the Future of On-Device AI



Ziad Asghar

Vice President of Product Management
Qualcomm Technologies Inc.

Powering the Connected Intelligent Edge and the Future of On-Device AI



Ziad Asghar

Vice President of Product Management
Qualcomm Technologies Inc.

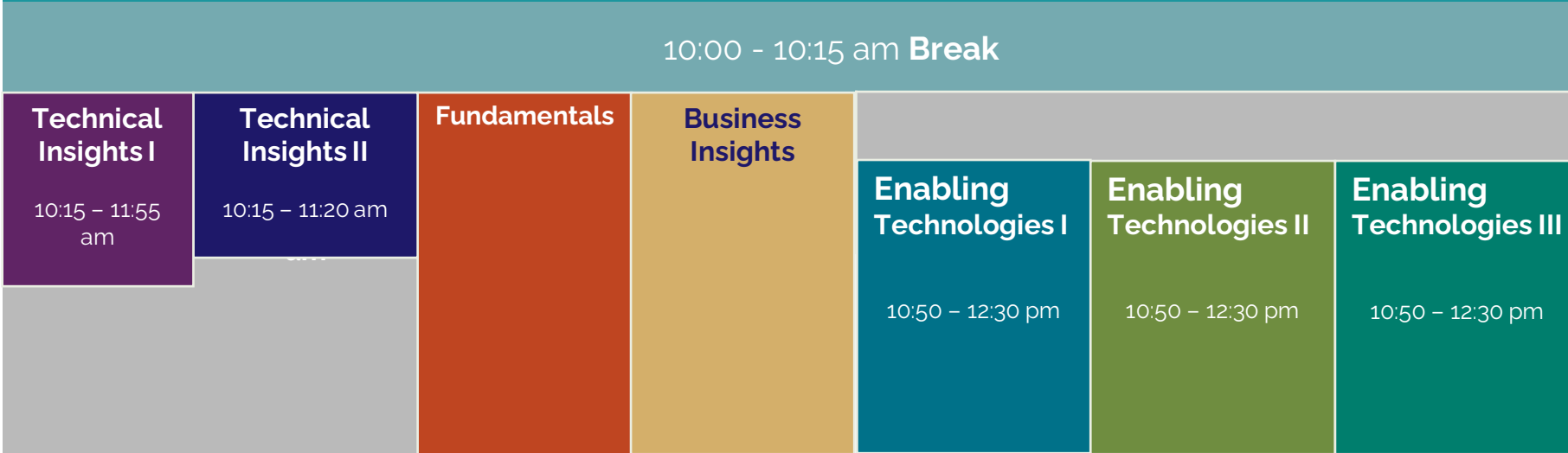
Wednesday Schedule



9:00 - 10:00 am

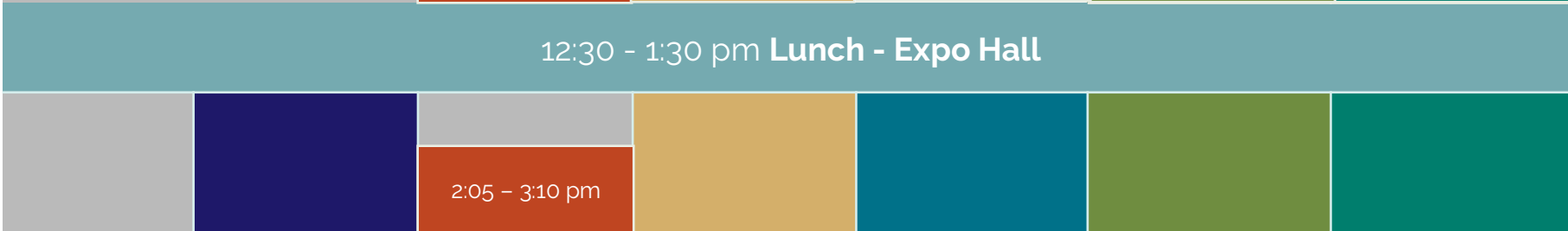
INTRODUCTION: A View from the Summit (Part 2) – Jeff Bier
GENERAL SESSION: Powering the Connected Intelligent Edge and the Future of On-Device AI – Ziad Asghar

10:15 am - 12:30 pm



10:00 am - 6:00 pm
Technology Exhibits
 Expo Hall

1:30 - 3:10 pm



4:15 - 6:00 pm

