

Advanced
Technology
Labs – Brasil

Challenges in real-time event detection in video

Leonardo Nunes

lnunes@microsoft.com



Objective

- Motivation on real-time event detection in video
- Challenges when analyzing real-world data
- Some lessons learned
- Future research directions

Advanced Technology Lab, Brazil



Motivation



There are over 245 million surveillance cameras around the world¹ (and growing), a market of nearly US\$38 Billion/year



Real-time visual monitoring is not practical, so cameras are almost always only used for after-the-fact analysis



Recent advances in image understanding are finally making truly smart automated VA possible and economically viable

Video analytics market will
*grow from US\$1.5B to
>US\$4B/year in 5 years
(>20% CAGR)²*

Event Detection

- Event/Action recognition:
 - Triggered when something of interest happens in a video
 - Caused by either a human, vehicle, or an object
- Major applications:
 - Surveillance
 - Mobility
 - Retail

Constraints



Real-time



Cloud-based



COTS sensors

Microsoft Cognitive Services

Cognitive Services API Collection



Vision

Computer Vision | Emotion | Face | Video



Speech

Computer Recognition | Speaker Recognition Speech | Translator



Language

Bing Spell Check | Language Understanding
Linguistic Analysis | Text Analytics | Web Language
Model



Knowledge

Academic Knowledge | Entity Linking Knowledge
Exploration | Recommendations



Search

Bing Autosuggest | Bing Image Search | Bing News
Search Bing Video Search | Bing Web Search

Applications

Surveillance Scenarios:

Traffic Event Detection
Person-of-Interest Detection
Suspicious Activity Detection
Asset Protection
Unattended Object Detection
Perimeter Intrusion
Retail Loss Prevention
Crowd Management

Additional Scenarios:

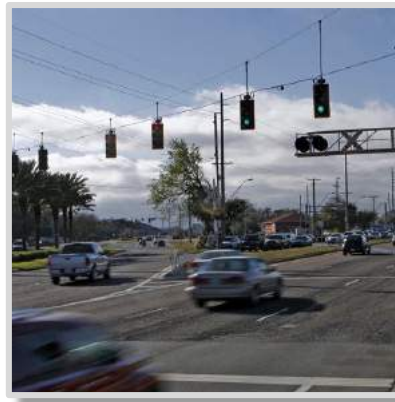
Retail Analytics
Natural Disaster Detection
Automotive Safety
Traffic Flow Measurement
Industrial Safety
Fire/Smoke/Flood Detection
Parking Analytics
Satisfaction Assessment
Vandalism Detection
Customer Recognition
People Counting
Home Automation
Accident Detection
Automated Media Indexing

Example Applications



People

- Person identification
- Attribute description
- Tracking



Vehicles

- Vehicle identification
- Model and maker ID
- Speed and trajectory estimation



Objects

- Classification

Challenges

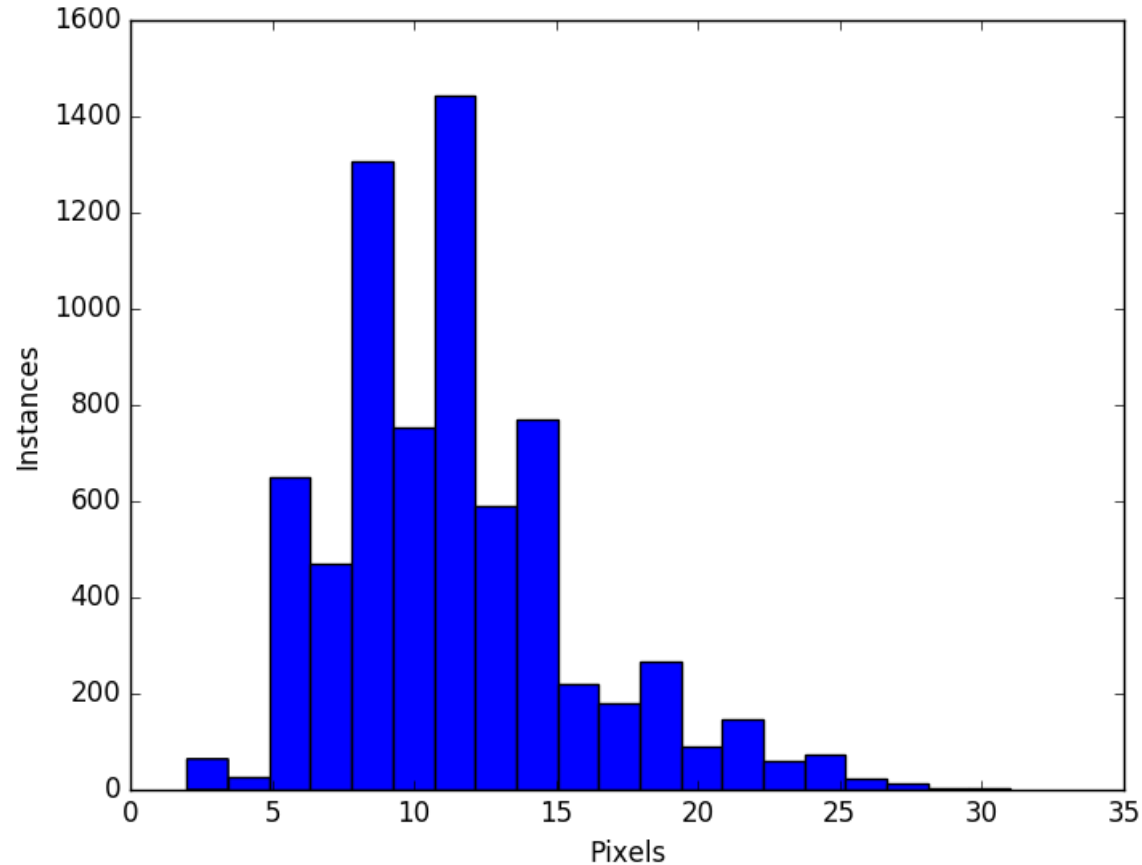


Video Quality

- Low-quality sensors
- Severe compression artifacts
- Video information is not reliable!
 - Repeated frames
 - Upsampled images



Small Targets



- Usually small targets
- Example:
 - Motorcycles
 - Single camera
- Off the shelf detectors might not work

Enviromental Conditions



Events of Interest are Rare

- Events of interest are rare
 - E.g.: crimes, accidents, person identification
- No training example available
- False alert rate vs Probability of detection
- Example (mobility):
 - Jaywalking: 3 events/hour
 - Motorcycle stopped: 0.2 events/hour

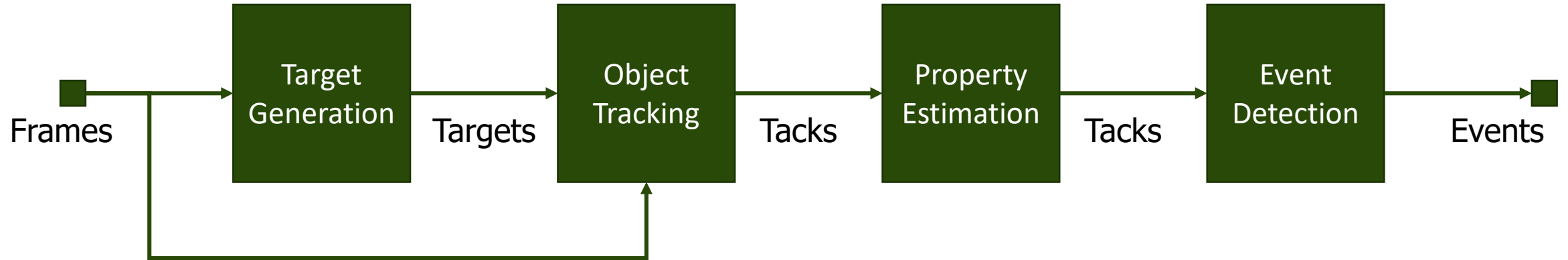
Infrastructure Issues

- Variable bit-rate
 - As low as 200 kbps
- Network issues
- Camera movement
- Variable CPU time
- ... all leading to lost frames

Some Lessons Learned



Pipeline System



Reliable Detectors

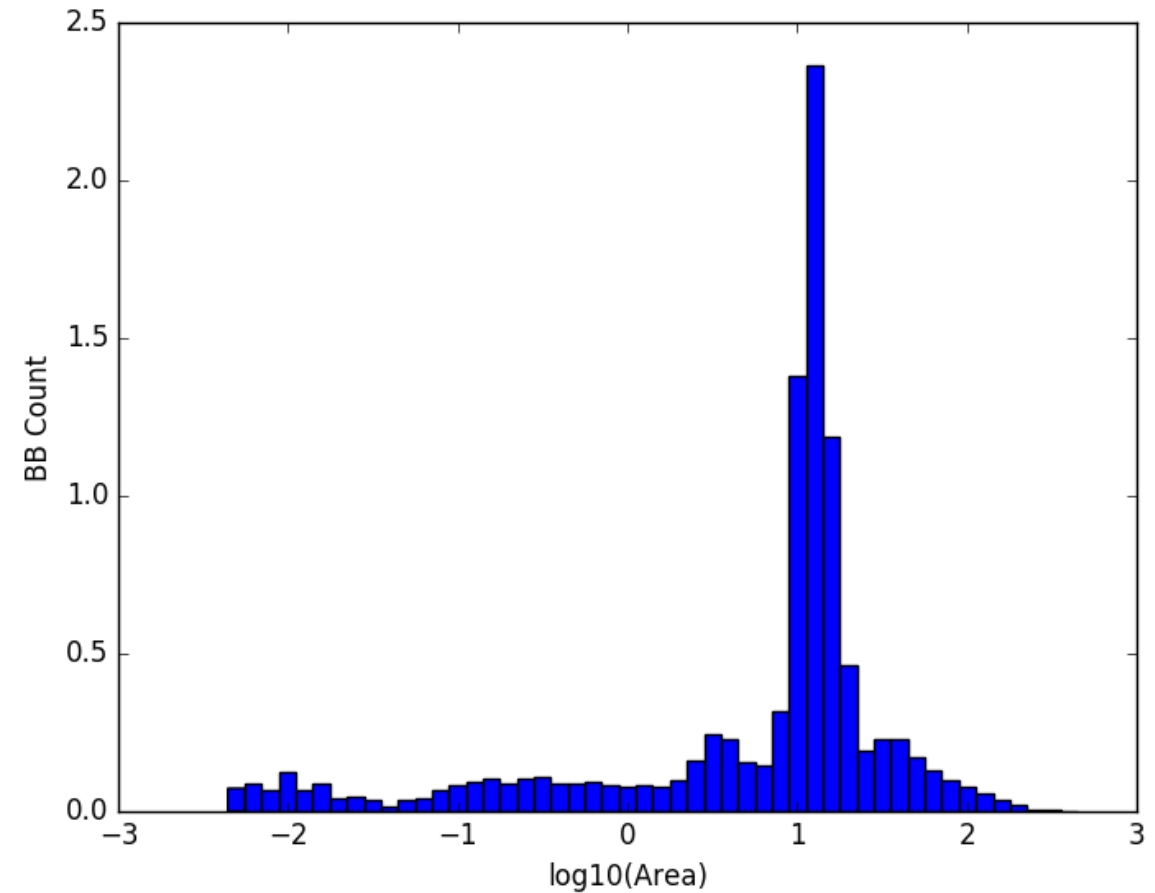
- Low-quality / small objects
 - Robust background subtraction algorithm
- Better quality / larger objects
 - Deep object recognition algorithms
- Most expensive component!
 - Not executed every frame: new targets are infrequent
 - Only when and where there is motion

Tracking by Detection

- Object detector executed:
 - Old tracks are updated with new targets
 - Similarity: distance and appearance
- Object detector not executed:
 - Tracks updated using Kernelized Correlation Filter
- Net effect: not every pixel is analyzed every frame

Ancillary information helps

- Perspective calibration
- Unsupervised learning of target sizes



Need to monitor network/hardware conditions

- Scene conditions detection:
 - Anomalous luminosity
 - Number of discarded frames
 - Camera movement
- Need paths to halt and re-initialize different components

Future



Video vs Image Algorithms

- When to use image vs video detectors?
- Deep object recognizers for video
- Improvements for small objects
- End to end learning

Unsupervised adaptation

- When to run target detectors?
- How to adapt detectors to camera?
- How to adapt tracker parameters?
- How to integrate deep architectures into the pipeline?

Summary

- Lots of challenges
- Specific tools, models for different tasks
- Lack of data, heuristics are still needed
- Unsupervised adaptation for different conditions



Thank you!

Leonardo Nunes

lnunes@microsoft.com