Media

Vision Framework Building on Core ML

Session 506

Brett Keating, Apple Manager Frank Doepke, He who wires things together

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#WWDC17



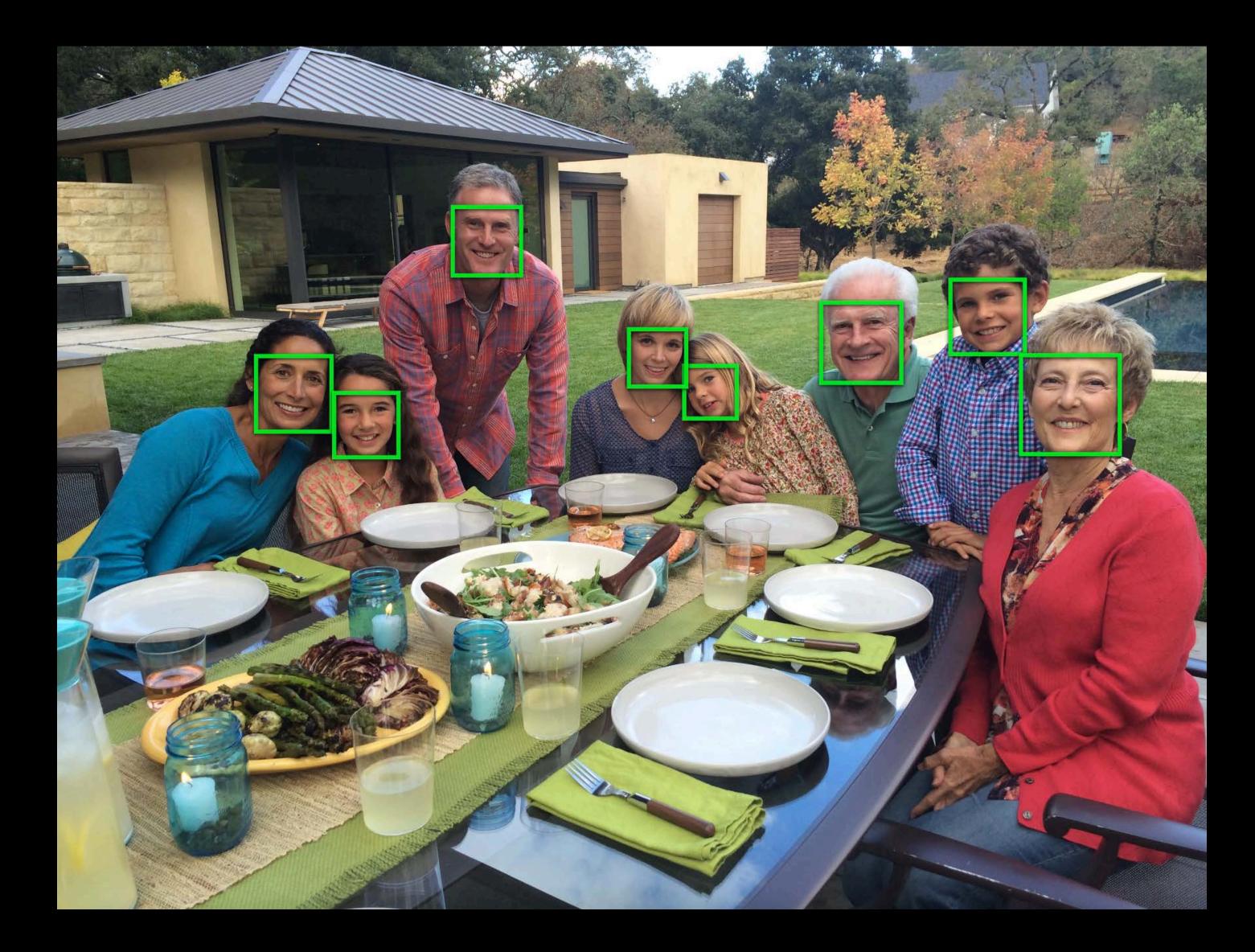
What Can Vision Do Vision Concepts The Code

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What You Can Do with Vision

Face Detection



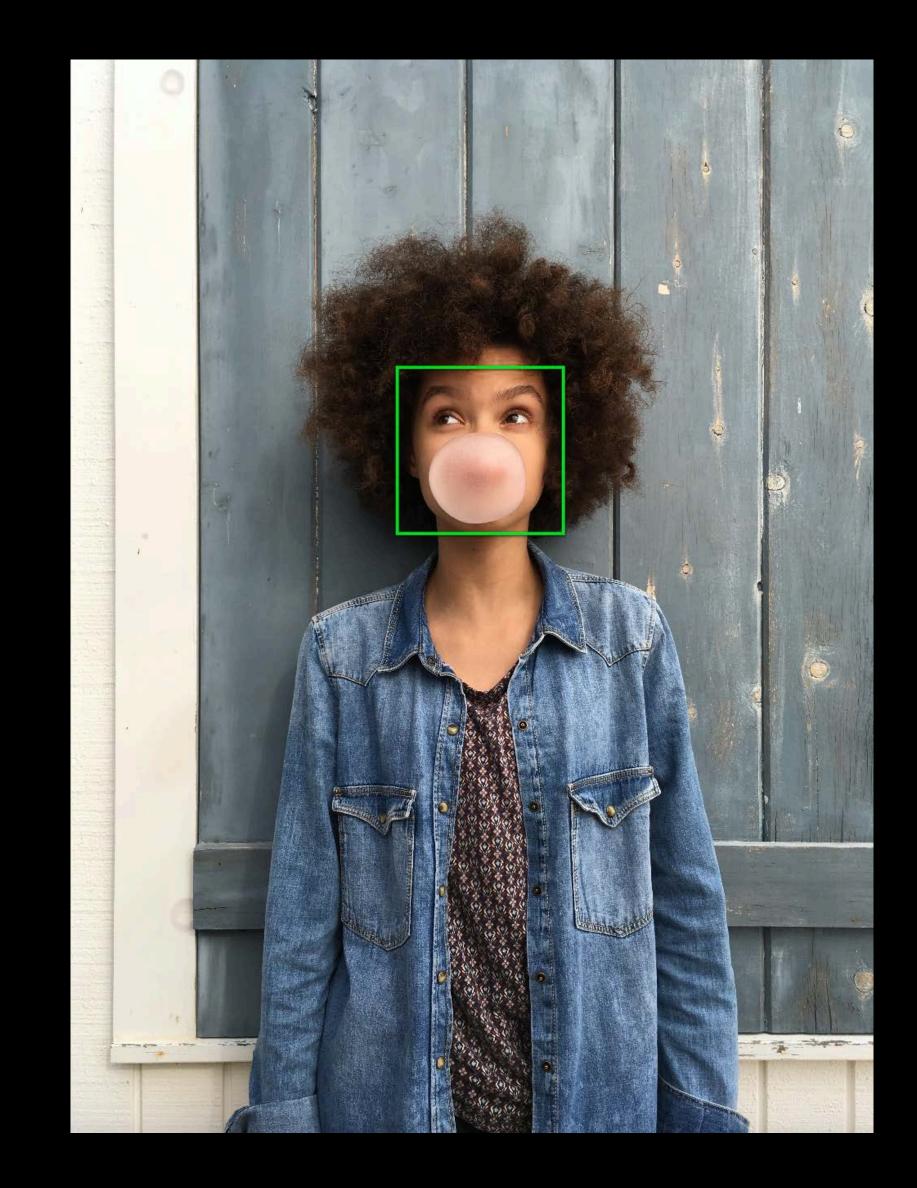
Face Detection: Small Faces



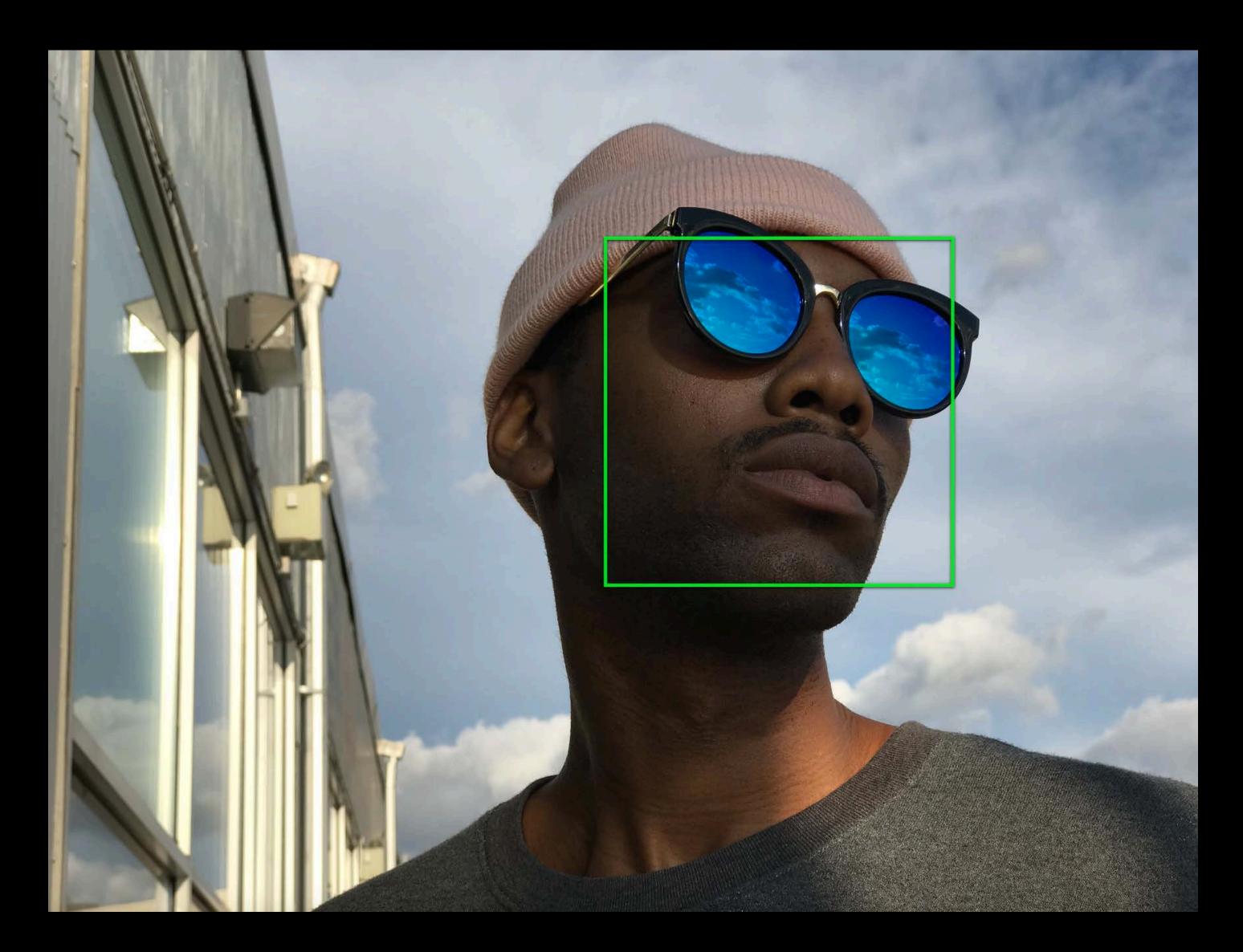
Face Detection: Strong Profiles

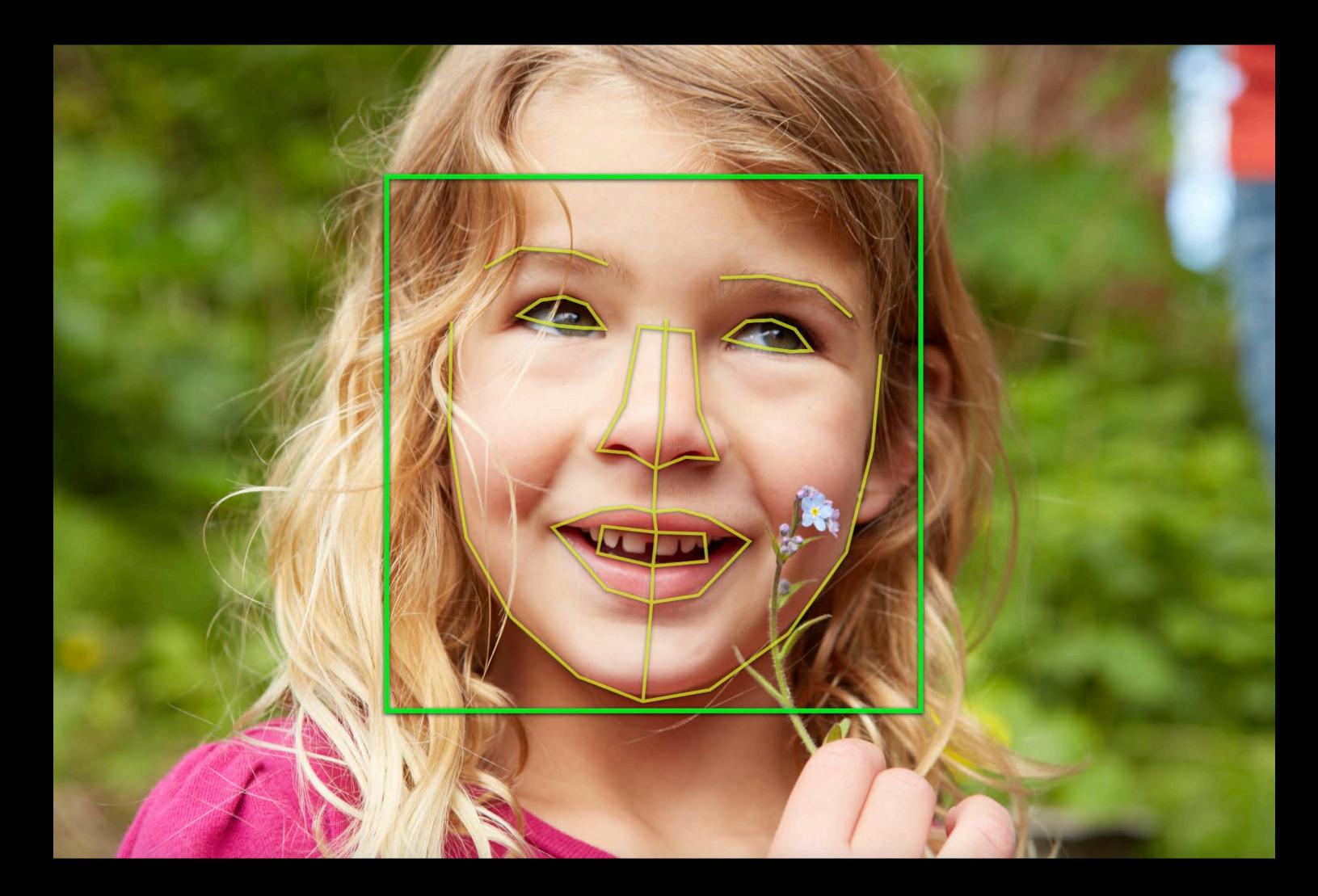


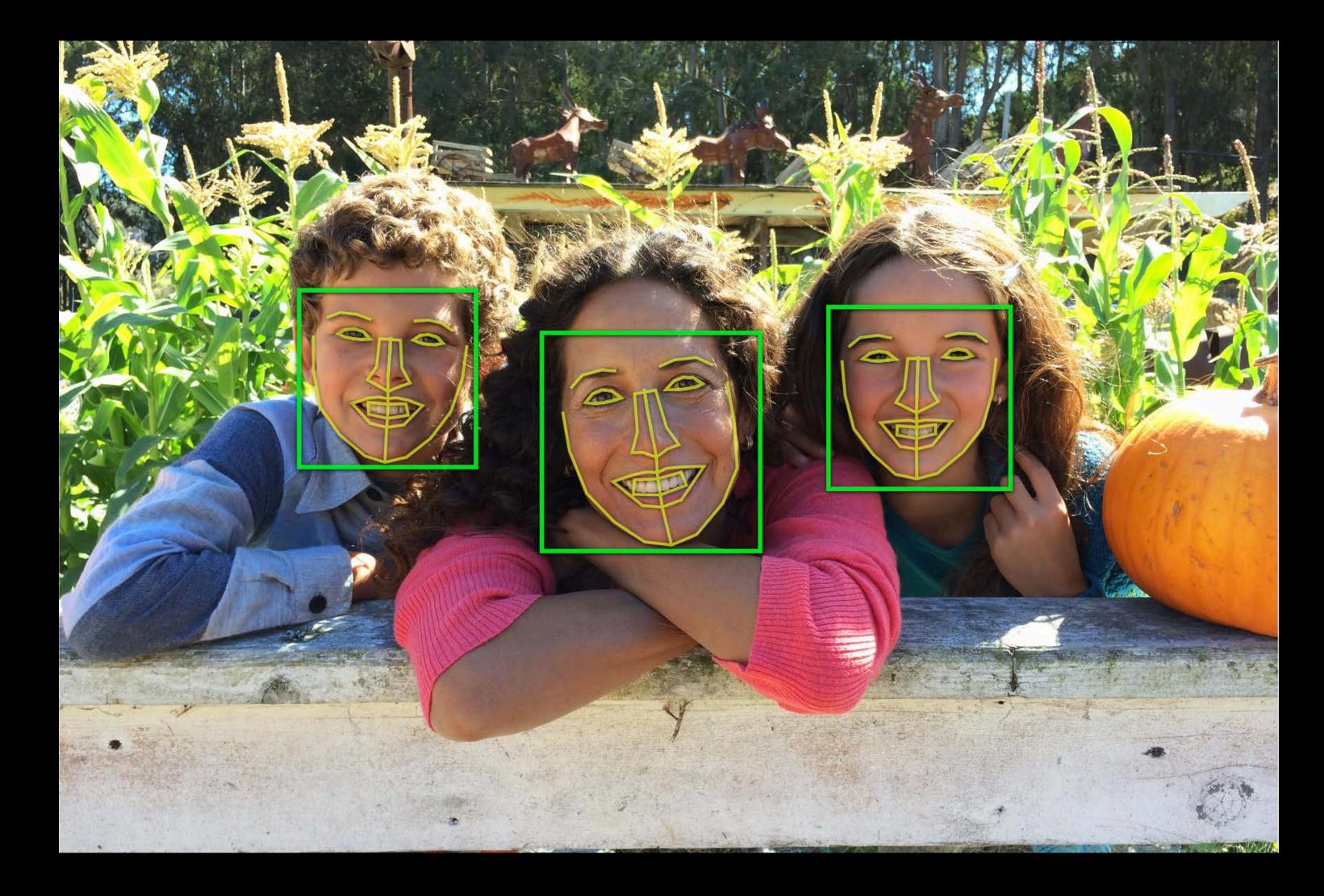
Face Detection: Partially Occluded



Face Detection: Hats and Glasses







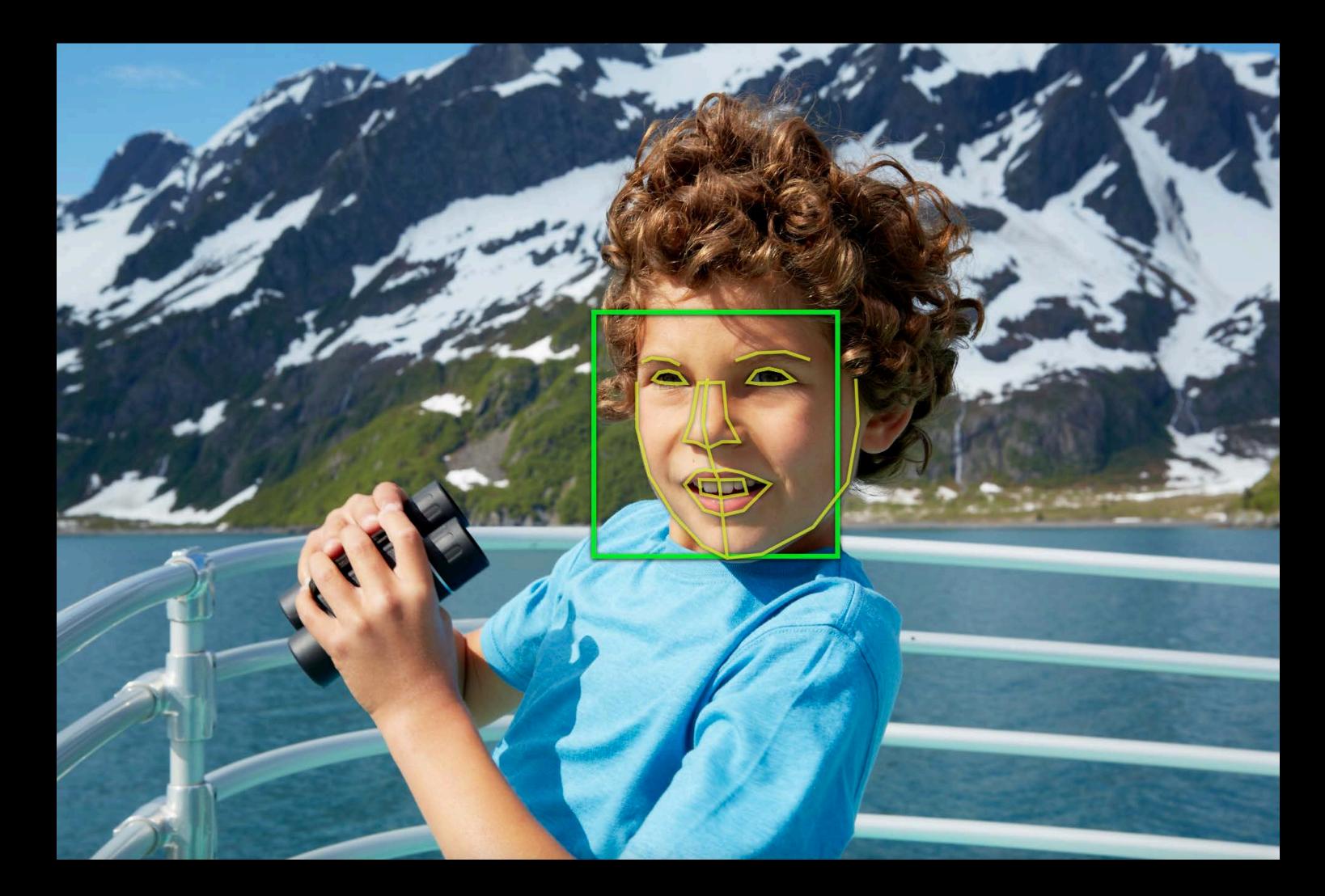


Image Registration

Image Registration

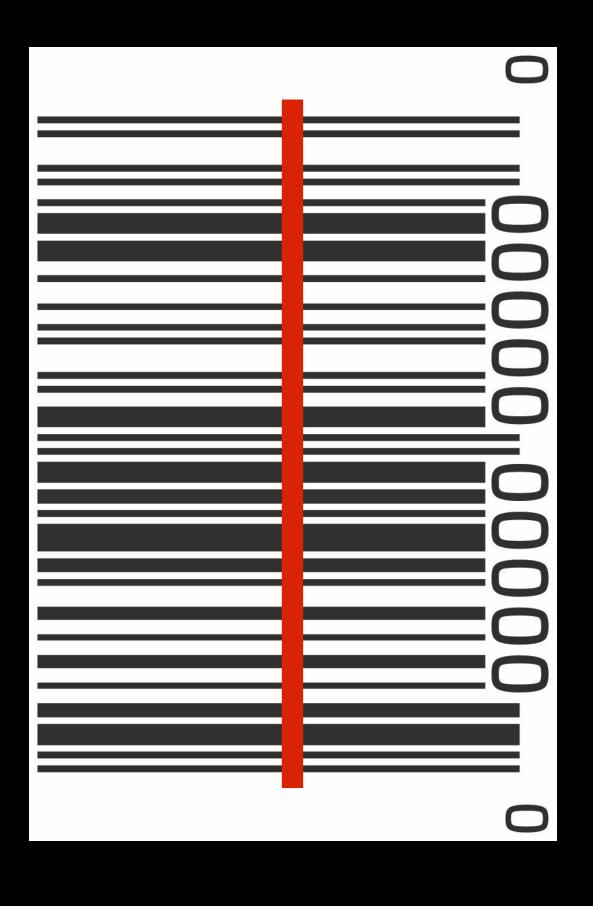


Rectangle Detection



Barcode Detection





Text Detection



Object Tracking For faces, rectangles, and general templates



Integration with Core ML

Significant advances in Computer Vision through Machine Learning

Core ML provides native acceleration for custom models

Vision provides the imaging pipeline to support Core ML models

Core ML and Natural Language Processing Lab

Core ML and Natural Language Processing Lab



Classification: Ballpoint, ballpoint pen, ballpen

Confidence: 0.86

Technology Lab D

Thur 11:00AM-3:30PM

Technology Lab D

Fri 1:50PM-4:00PM

High-level on-device solutions to Computer Vision problems through one simple API

You don't have to be a Computer Vision expert

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- "I just want to know where the faces are"
- Handles the complexity for you

- You don't have to be a Computer Vision expert
- "I just want to know where the faces are"
- Handles the complexity for you
- Traditional and deep learning algorithms

Privacy

Images and video stay on device

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Cost

- No usage fees
- No data transfer

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Real-time use cases

No latency, fast execution

Vision Concepts

Frank Doepke, He who wires things together

The Asks

The Machinery

The Results

Requests

VNDetectBarcodesRequest VNDetectFaceLandmarksRequest VNDetectFaceRectanglesRequest

The Machinery

The Results

Requests

VNDetectBarcodesRequest VNDetectFaceLandmarksRequest VNDetectFaceRectanglesRequest





RequestHandler



The Results

Requests

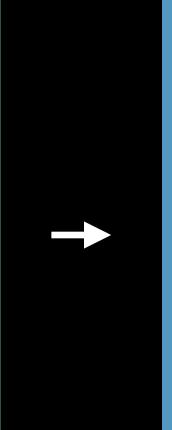
VNDetectBarcodesRequest VNDetectFaceLandmarksRequest VNDetectFaceRectanglesRequest





RequestHandler





Observations

VNClassificationObservation VNDetectedObjectObservation VNFaceObservation

The Asks

The Machinery

The Results

Requests

VNTrackObjectRequest VNTrackRectangleRequest VNImageRegistrationRequest



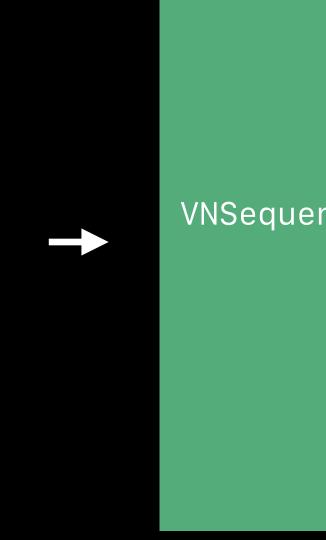
The Machinery

The Results

Requests

VNTrackObjectRequest VNTrackRectangleRequest VNImageRegistrationRequest





RequestHandler

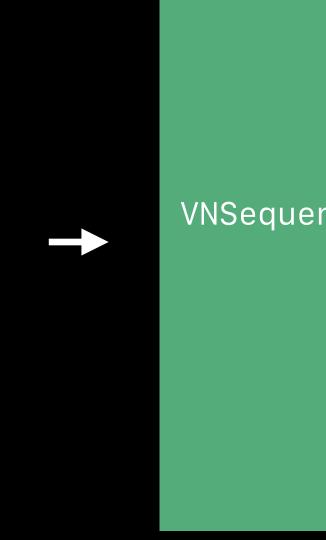
VNSequenceRequestHandler

The Results

Requests

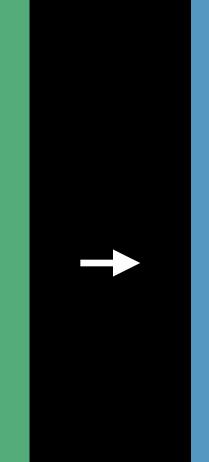
VNTrackObjectRequest VNTrackRectangleRequest VNImageRegistrationRequest





RequestHandler

VNSequenceRequestHandler



Observations

VNDetectedObjectObservation VNRectangleObservation VNImageAlignmentObservation

Image Request Handler

- For interactive exploration of an image
- Holds on to the image for its lifecycle
- Allows optimization of various requests performed on an image

Sequence Request Handler

For anything that looks at images in a sequence like tracking Does not optimize for multiple requests on an image

Putting It into Code



// Create request handler

let myRequestHandler = VNImageRequestHandler(url: fileURL, options: [:])

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// send the requests to the request handler
myRequestHandler.perform([faceDetectionRequest])

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let myRequestHandler = VNImageRequestHandler(url: fileURL, options: [:])

// send the requests to the request handler
myRequestHandler.perform([faceDetectionRequest])

}

// Do we have a face
for observation in faceDetectionRequest.results as! [VNFaceObservation] {
 /// do something

// Create request handler
let myRequestHandler = VNImageRequestHandler(url: fileURL, options: [:])

// send the requests to the request handler
myRequestHandler.perform([faceDetectionRequest])

// Start the tracking with an observation let observations = detectionRequest.results as! [VNDetectedObjectObservation] let objectsToTrack = observations.map { VNTrackObjectRequest(detectedObjectObservation: \$0) }

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// Run the requests

requestHandler.perform(objectsToTrack, on: pixelBuffer)

// Start the tracking with an observation let observations = detectionRequest.results as! [VNDetectedObjectObservation] let objectsToTrack = observations.map { VNTrackObjectRequest(detectedObjectObservation: \$0) }

// Run the requests requestHandler.perform(objectsToTrack, on: pixelBuffer)

// Lets look at the results for request in objectsToTrack for observation in request.results as! [VNDetectedObjectObservation]

Best Practices

Which image type is right for me?

Which image type is right for me? What am I going to do with the image?

Which image type is right for me? What am I going to do with the image? What performance do I need or want?

Vision supports various image types

CVPixelBufferRef CGImageRef CIImage NSURL NSData

Vision supports various image types

- CVPixelBufferRef
- CGImageRef
- CIImage
- NSURL
- NSData

The image type to choose depends on where the image comes from

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CVPixelBufferRef CGImageRef

CIImage

NSURL

NSData

The image type to choose depends on where the image comes from You shouldn't have to pre-scale the image

Vision supports various image types

CVPixelBufferRef CGImageRef CIImage NSURL NSData

The image type to choose depends on where the image comes from You shouldn't have to pre-scale the image Make sure to pass in the EXIF orientation of the image

Everything Streaming

CVPixelBuffer

Comes from a CMSampleBuffer in the VideoDataOut of a camera stream

Also a good low-level format to provide image data in memory

Files from Disk or Web

URL for image files on disk NSData for images from the web Least amount of memory footprint Vision will do the scaling without reading the full image if possible EXIF Orientation is derived from the file if possible but can be overwritten

Core Image

Already using Core Image Preprocessing the image

Advances in Core Image: Filters, Metal, Vision, and More

Executive Ballroom

Thursday 1:50PM

Images Already Used in the UI

Use CGImage if the image was already used in the UI Ullmage and NSImage have accessors for CGImageRefs

What Am I Going to Do with the Image?

What Am I Going to Do with the Image?

Interactively explore the image

- Use VNImageRequestHandler and hold onto it
- Remember that the input image is immutable

old onto it mutable

What Am I Going to Do with the Image?

Interactively explore the image

- Use VNImageRequestHandler and hold onto it
- Remember that the input image is immutable

Tracking an observation

- Use VNSequenceRequestHandler
- Tracking state is kept in the VNSequenceRequestHandler
- Lifecycle of images is not tied to the life of the VNSequenceRequestHandler

old onto it mutable

What Performance Do I Need or Want?

Vision tasks can be time consuming and processing intensive

- Dispatch your work on a queue with appropriate QOS
- Use the completion handler to work with the results
- Completion handler is called on the same queue as the request

Yet Another Face Detector?

Vision uses deep learning for face detection

- Highest precision and recall
- Slower on older hardware in particular



Face Detector Landscape



Accuracy	Best
Processing time	Fast
Power usage	Good
Availability	iOS, macOS, tvOS

Core Image	AV Capture
Better	Good
Faster	Fastest
Better	Best
iOS, macOS, tvOS	iOS capture only

Core Image

Vision

Core Image



Vision



Core Image



Vision



CIDetector vs. Vision

CIDetector will remain as they are in Core Image New algorithms will be exposed through Vision Algorithm improvements will be made available in Vision

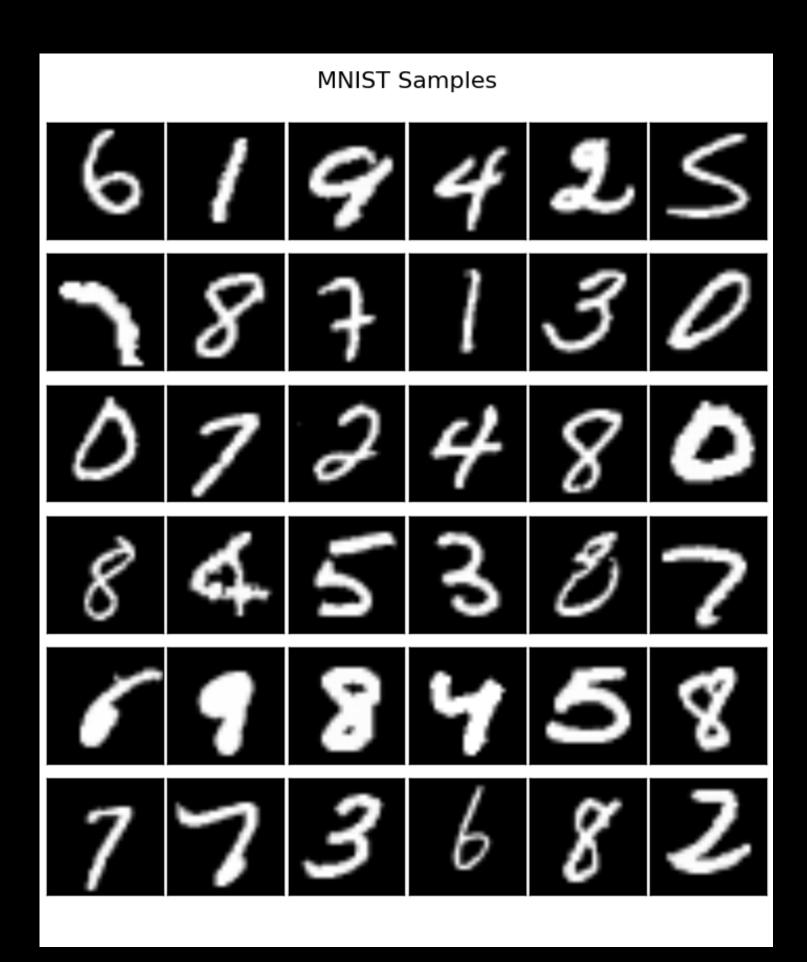
Show and Tell Part one

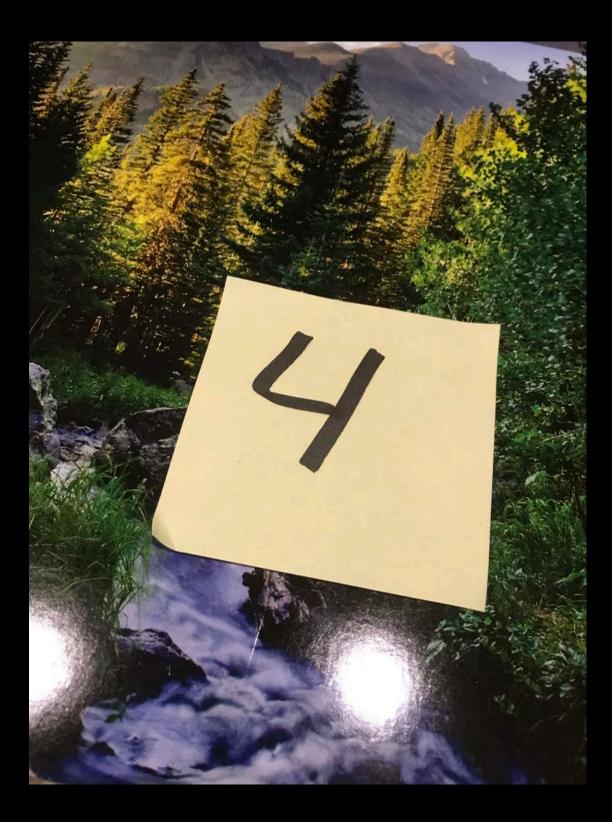
Show and Tell Part two

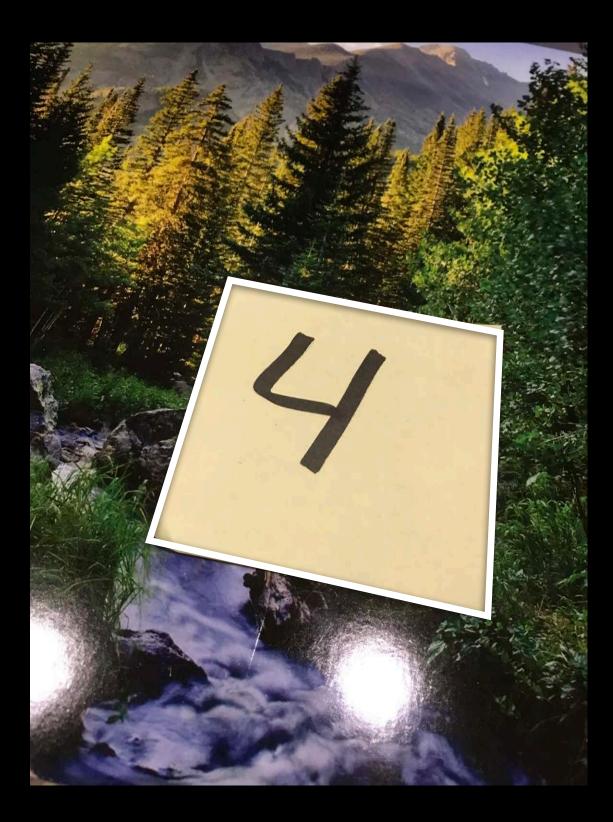
MNISTVision

Concepts to be covered

- Spin off requests from other requests
- Use Core Image for processing
- Use Core ML for machine learning

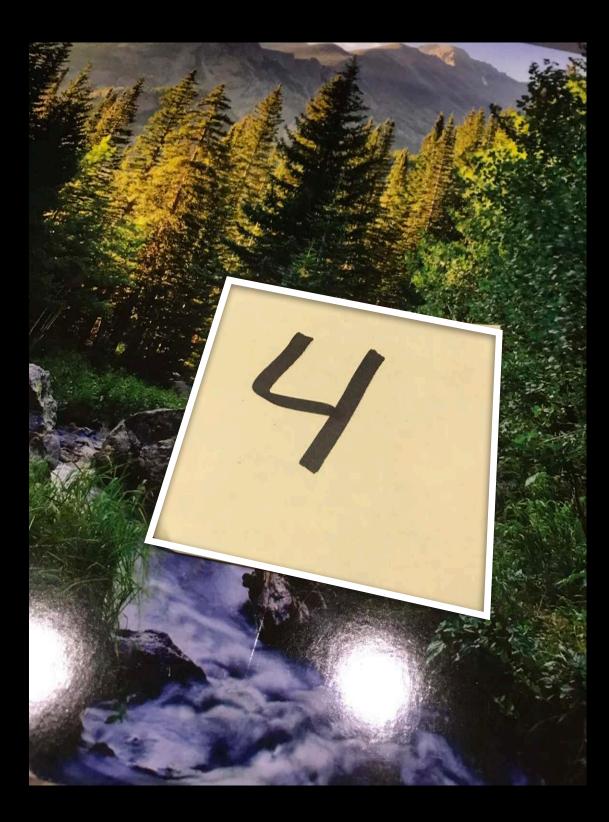


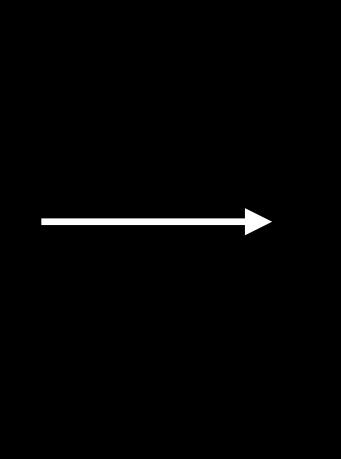


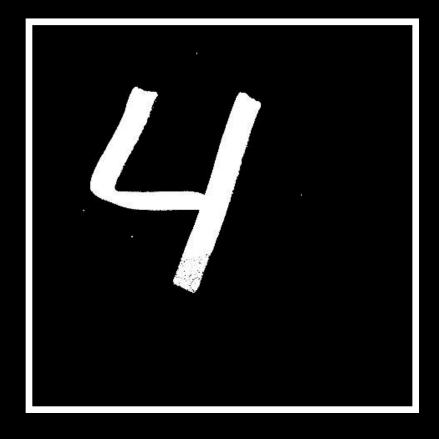


Step 1

Find sticky note using Rectangle Detector



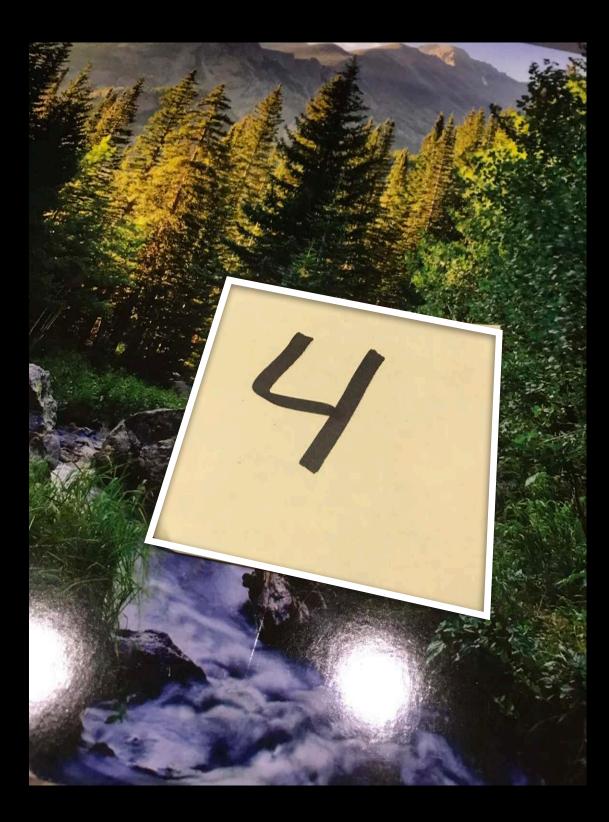


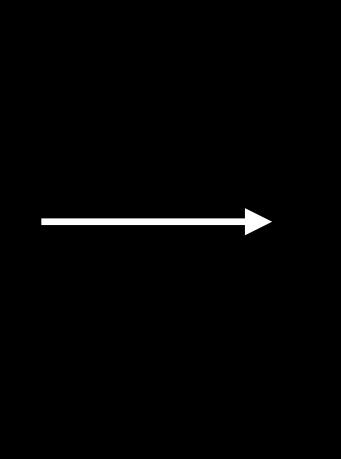


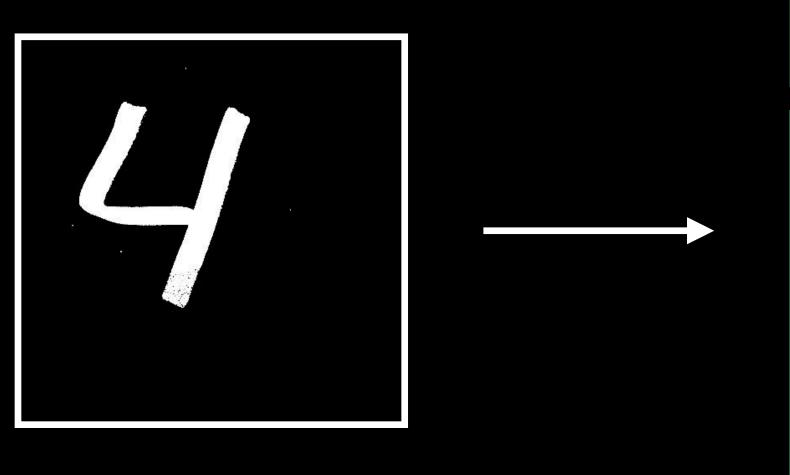
Step 1

Find sticky note using Rectangle Detector Use CI for perspective correction and image processing

Step 2







Step 1

Find sticky note using **Rectangle Detector**

Use CI for perspective correction and image processing

Confidence: 0.87

MNISTClassifier:

Step 2

Step 3

Run MNIST classifier on resulting CIImage





And that is Vision.framework

Summary

Vision is a new high-level framework for Computer Vision Various detectors and tracking through one consistent interface Integration with Core ML allows you to use custom models with ease

More Information

https://developer.apple.com/wwdc17/506

Related Sessions

Introducing Core ML

Photography Get Together

Core ML in Depth

Advances in Core Image: Filters, Metal, Vision, and More

	WWDC 2017
Technology Lab J	Wednesday 6:30PM
Hall 3	Thursday 9:00AM
Executive Ballroom	Thursday 1:50PM
	Hall 3



Core ML and Natural Language Processing Lab

AVFoundation Lab

Photos Editing and Core Image Lab

Vision Lab

Core ML and Natural Language Processing Lab

Technology Lab D	Thur 11:00AM–3:30PM
Technology Lab F	Thur 12:00PM–3:00PM
Technology Lab F	Thur 3:10PM–6:00PM
Technology Lab A	Fri 1:50PM-4:00PM
Technology Lab D	Fri 1:50PM-4:00PM

