

A Quintillion Live Pixels: The Challenge of Continuously Interpreting, Organizing, and Generating the World's Visual Information

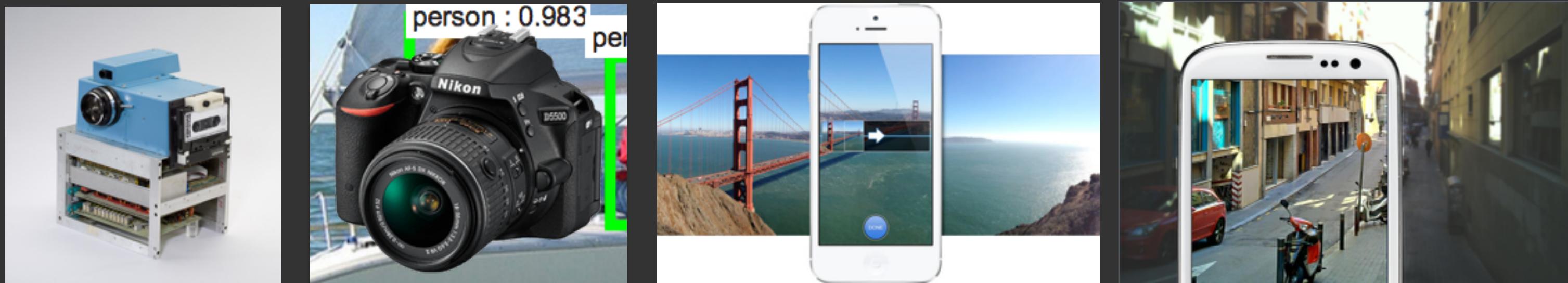
Kayvon Fatahalian
Carnegie Mellon University

Visual computing

2D/3D graphics



Image processing / computational photography



Computer vision (visual scene understanding)

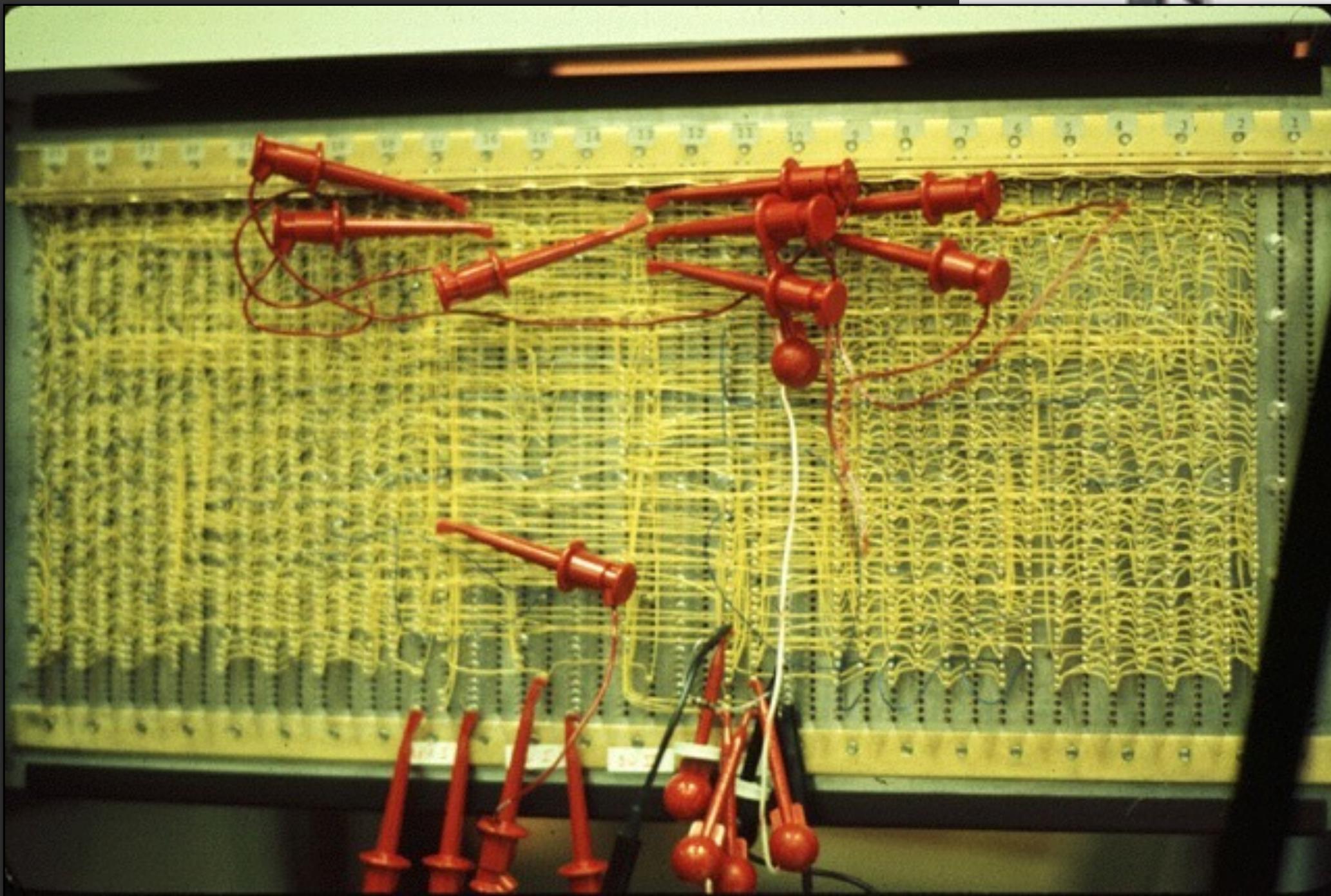




Ivan Sutherland's Sketchpad on MIT TX-2 (1962)

The frame buffer

Shoup's SuperPaint (PARC 1972-73)



16 2K shift registers (640 x 486 x 8 bits)

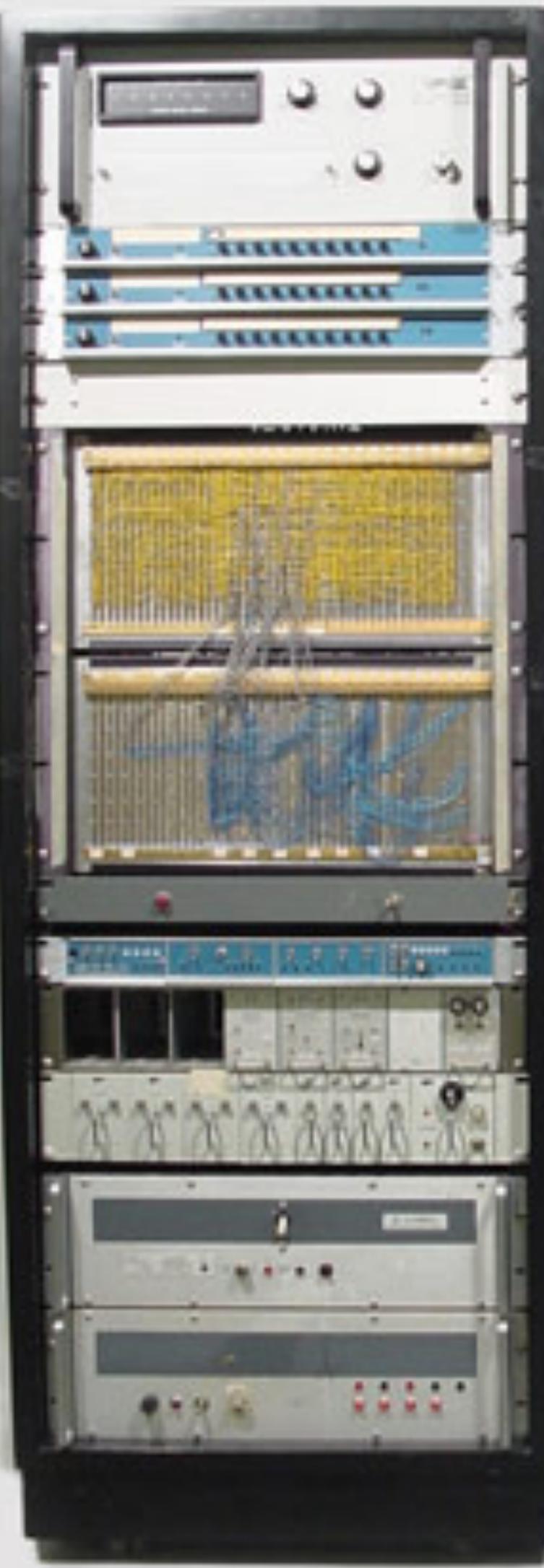


The frame buffer

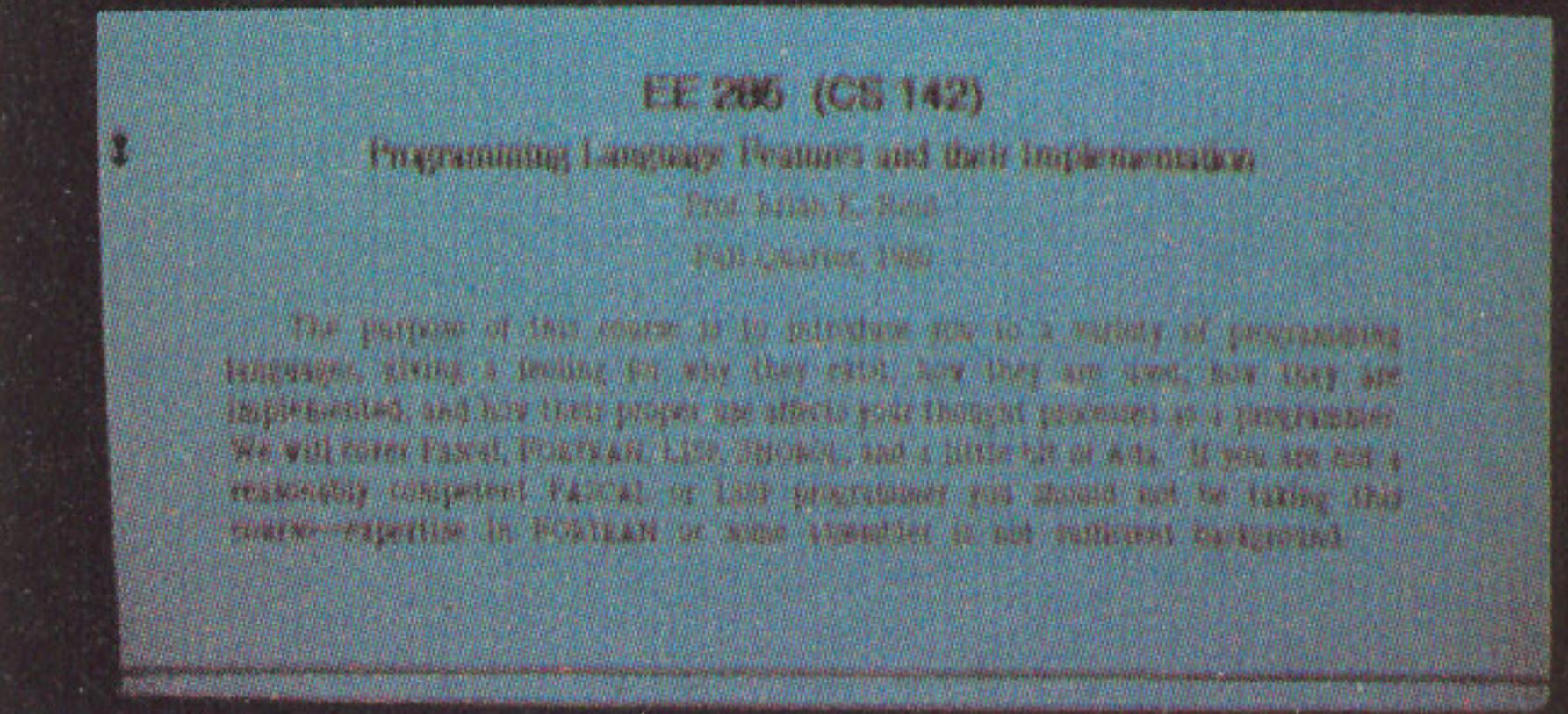
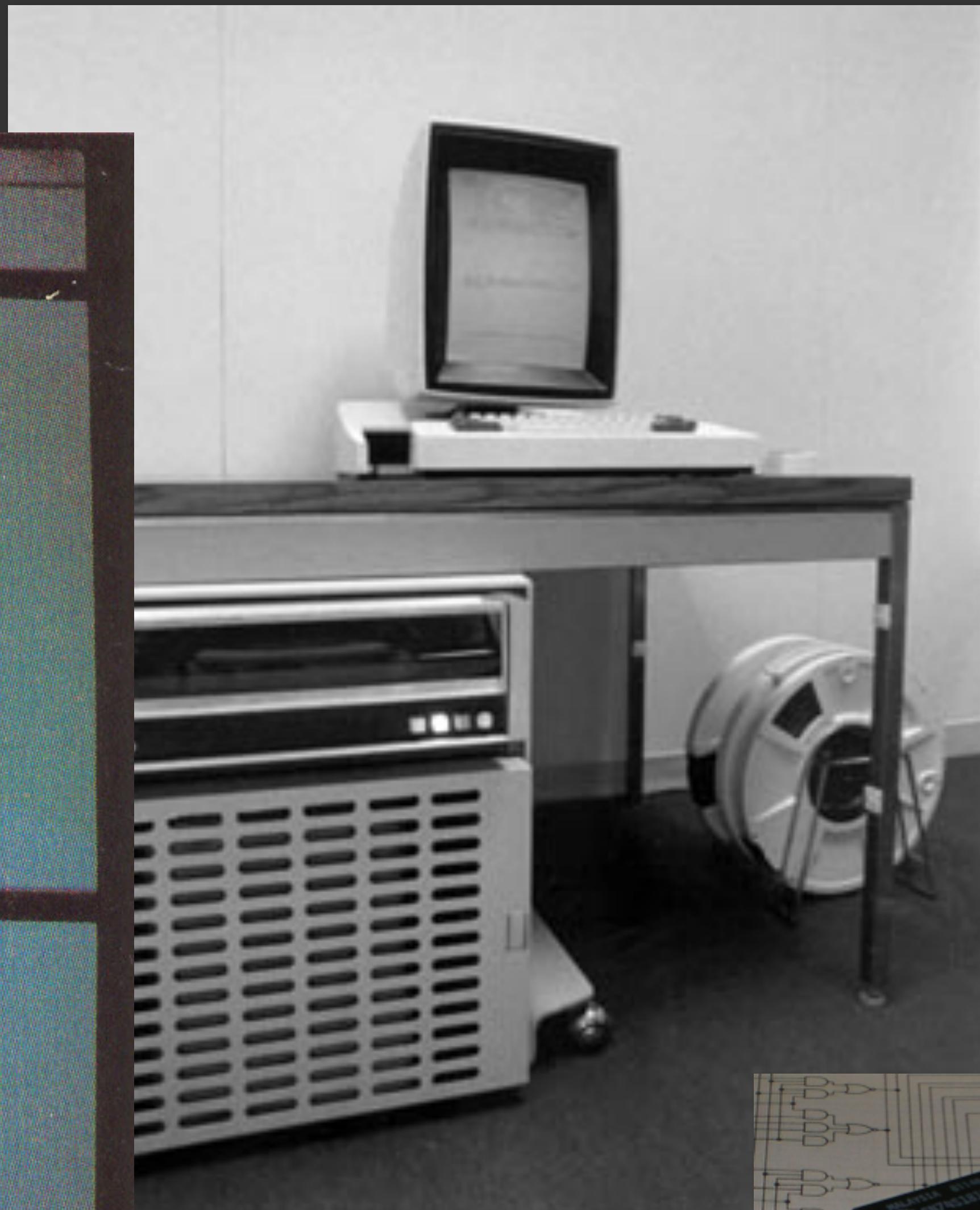
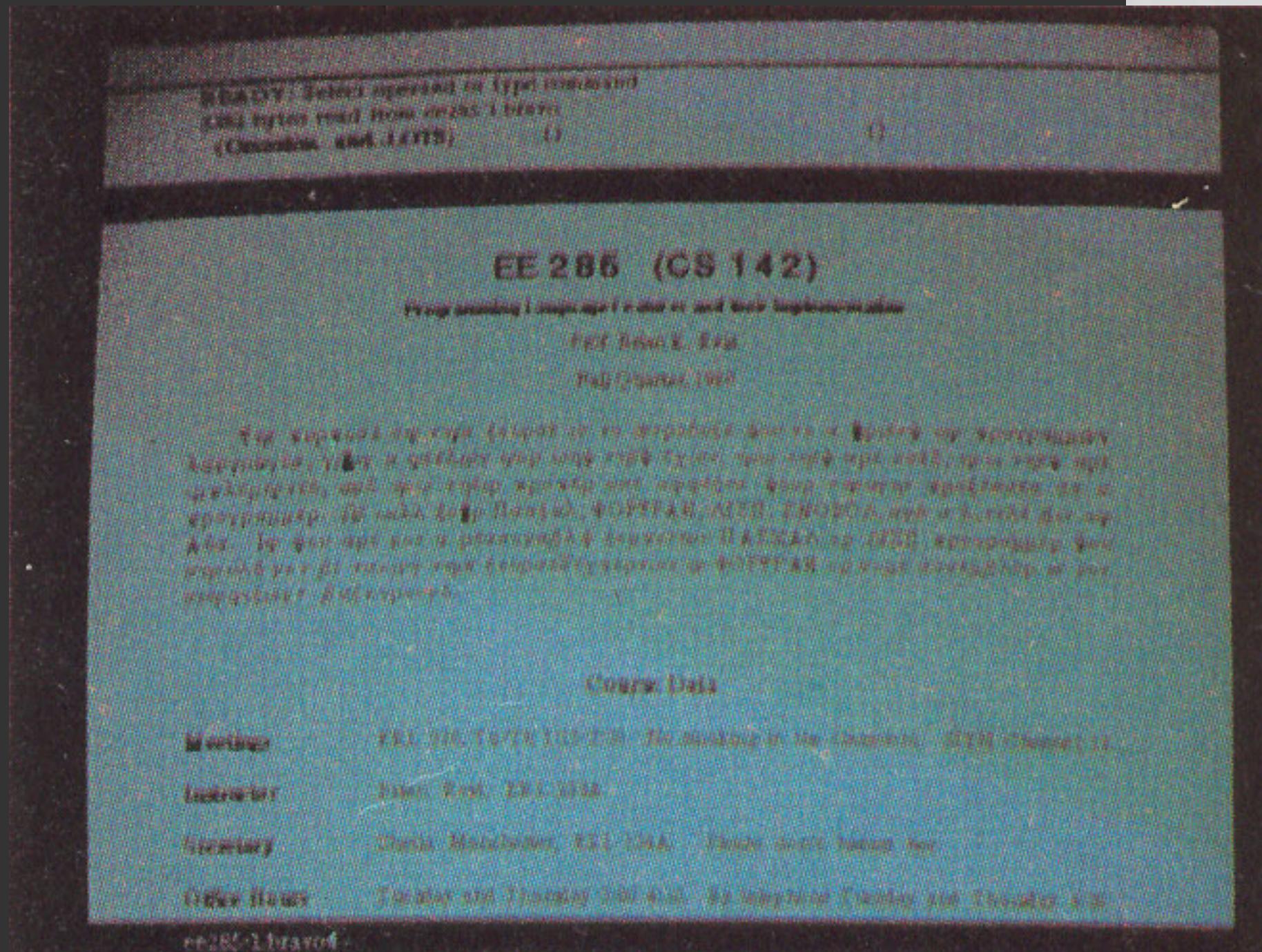
Shoup's SuperPaint (PARC 1972-73)



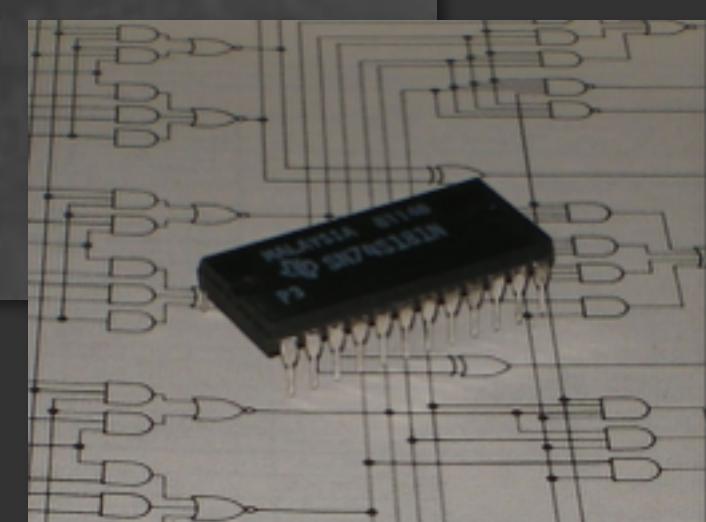
16 2K shift registers (640 x 486 x 8 bits)



Xerox Alto (1973)



Bravo (WYSIWYG)



TI 74181 ALU

Goal: render everything you've ever seen

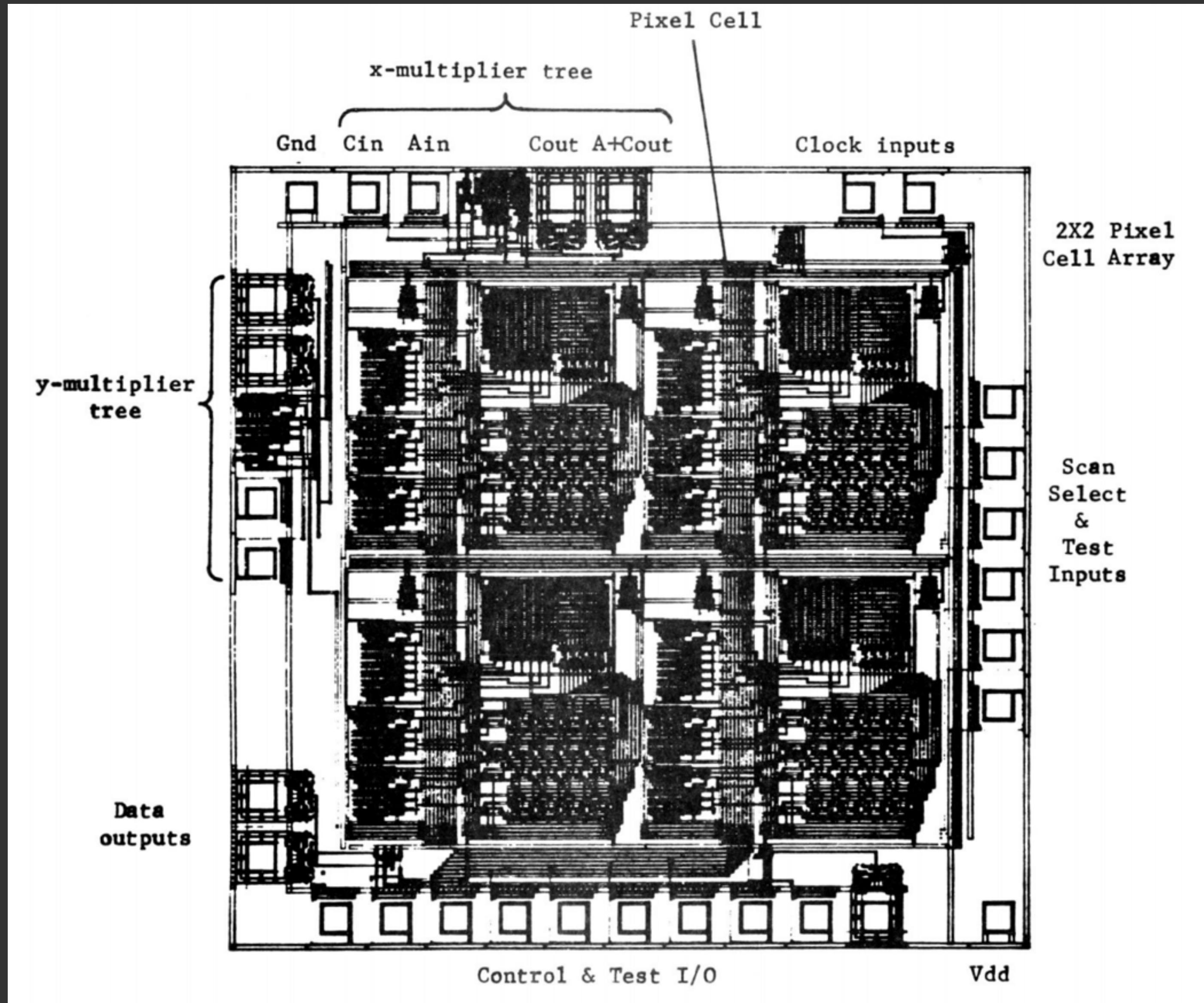
“Road to Pt. Reyes”
LucasFilm (1983)



Pixar's Toy Story (1995)



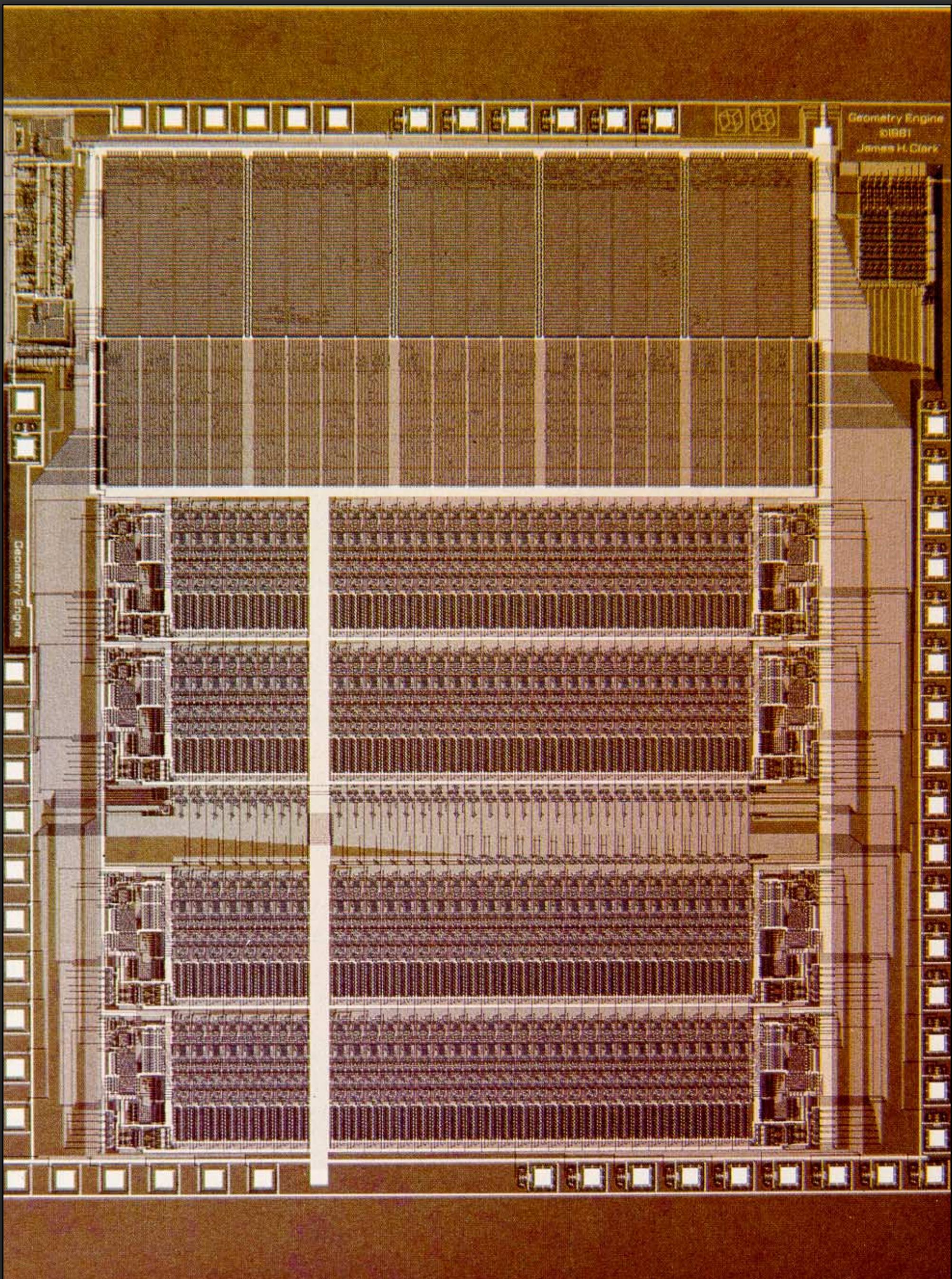
"We take an average of three hours to draw a single frame on the fastest computer money can buy."
- Steve Jobs

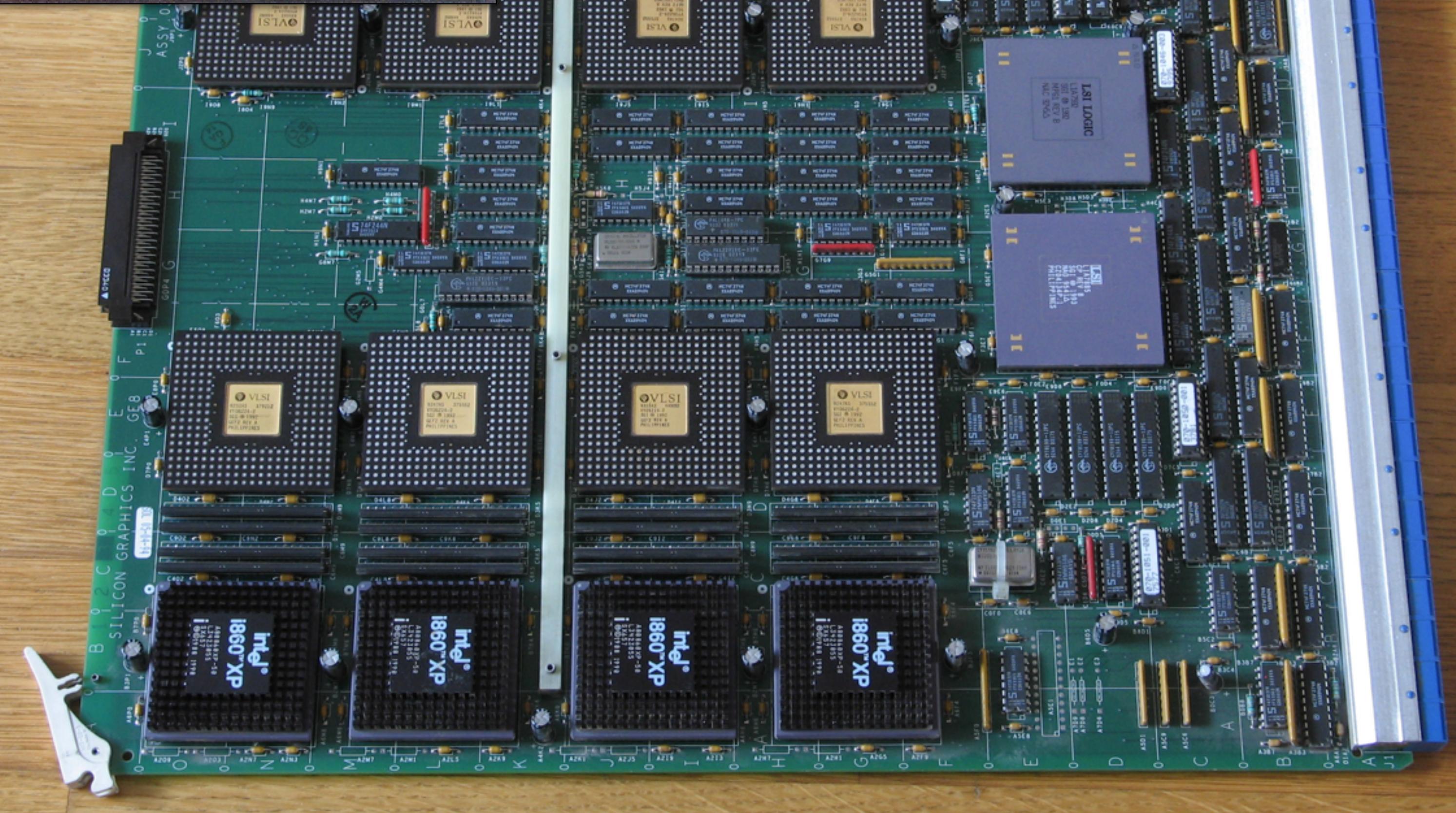


UNC Pixel Planes (1981), computation-enhanced frame buffer

Ed Clark's Geometry Engine (1982)

ASIC for geometric transforms
used in real-time graphics.



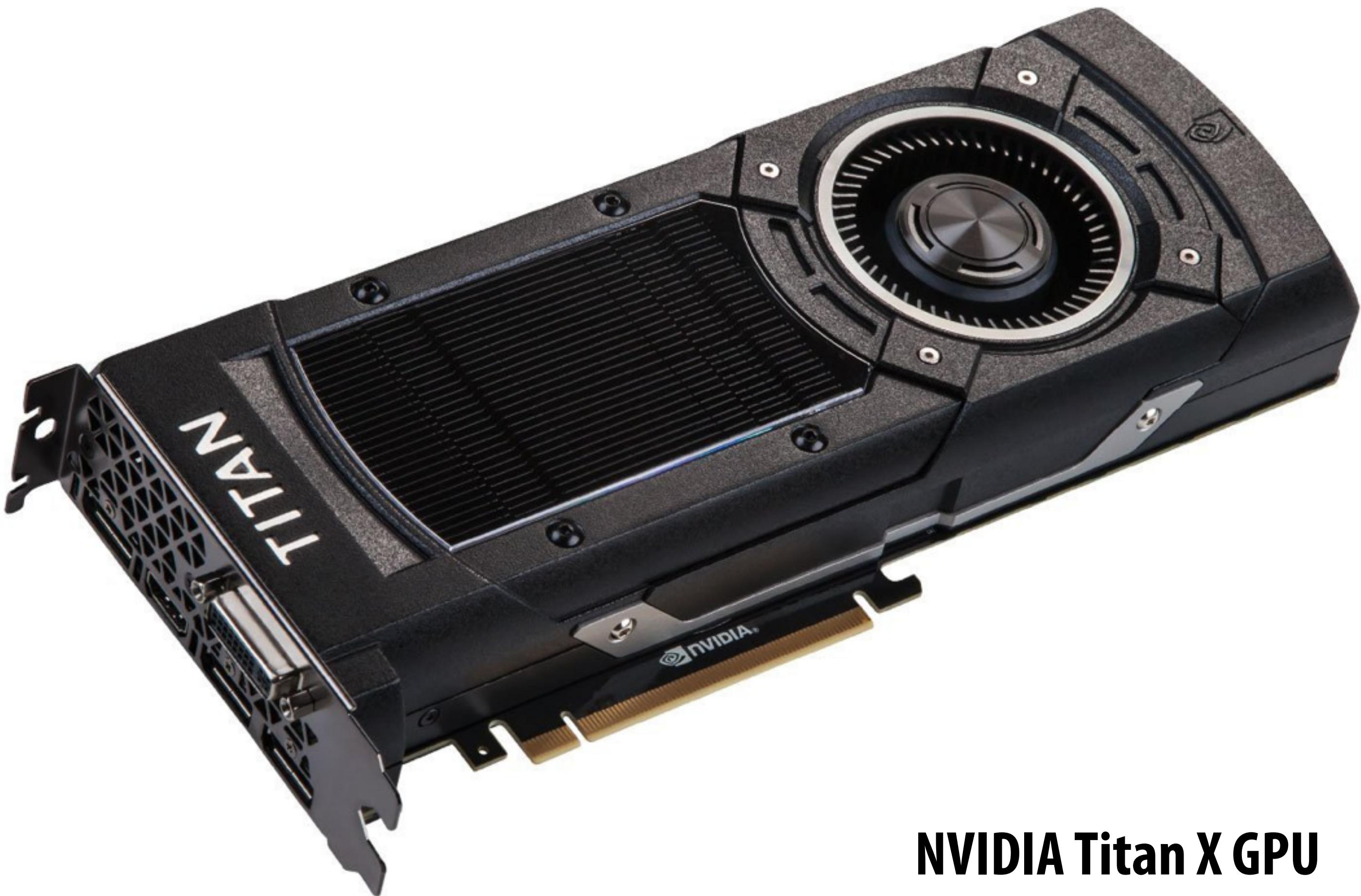


SGI RealityEngine GE8 board (1993)

Real-time (30 fps) on a NVIDIA Titan X



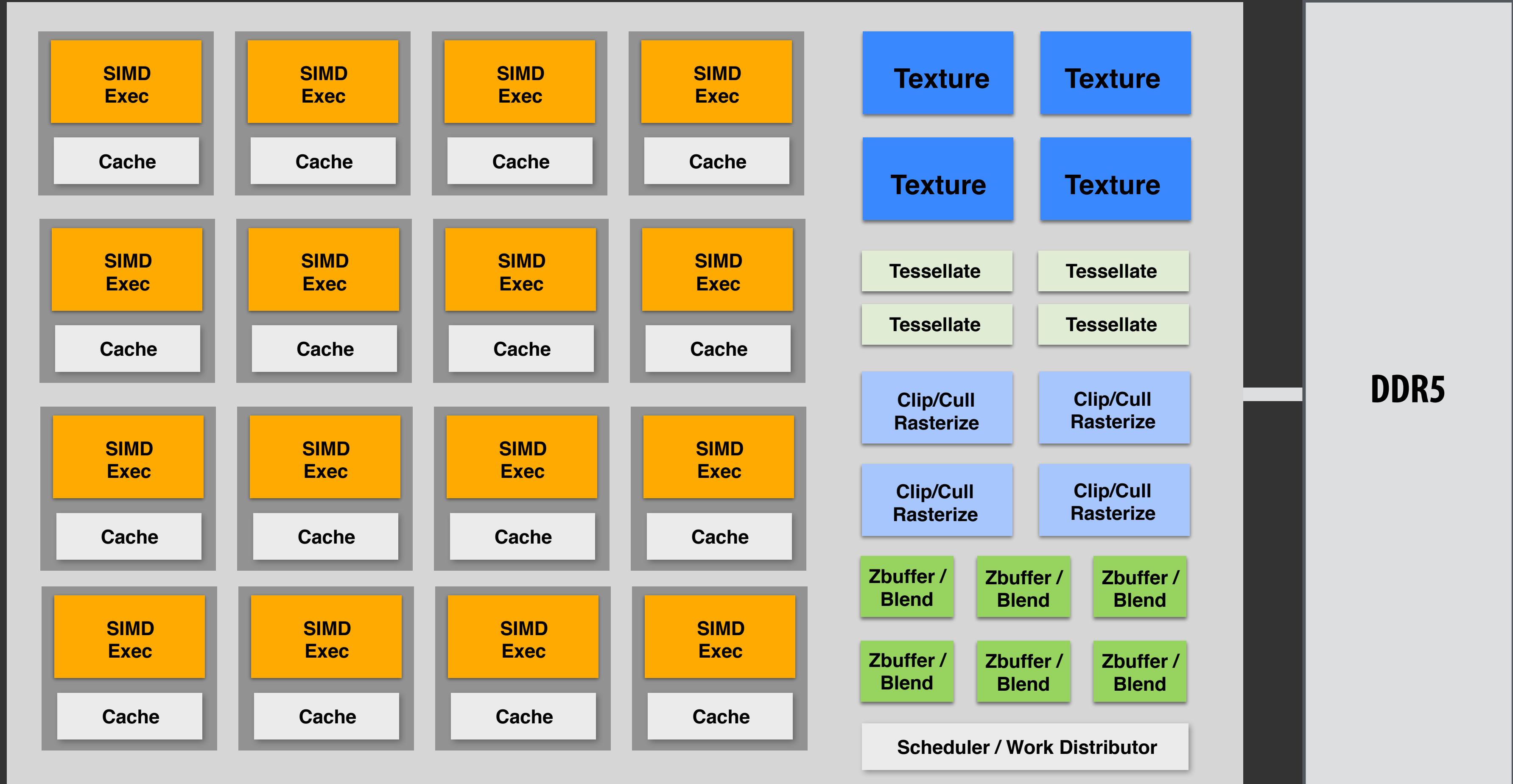
Unreal Engine Kite Demo (Epic Games 2015)



NVIDIA Titan X GPU
(~ 7 TFLOPs fp32)

Tesla generation NV chip ~ ASCI Red

Modern GPU: heterogeneous multi-core



Multi-threaded, SIMD cores

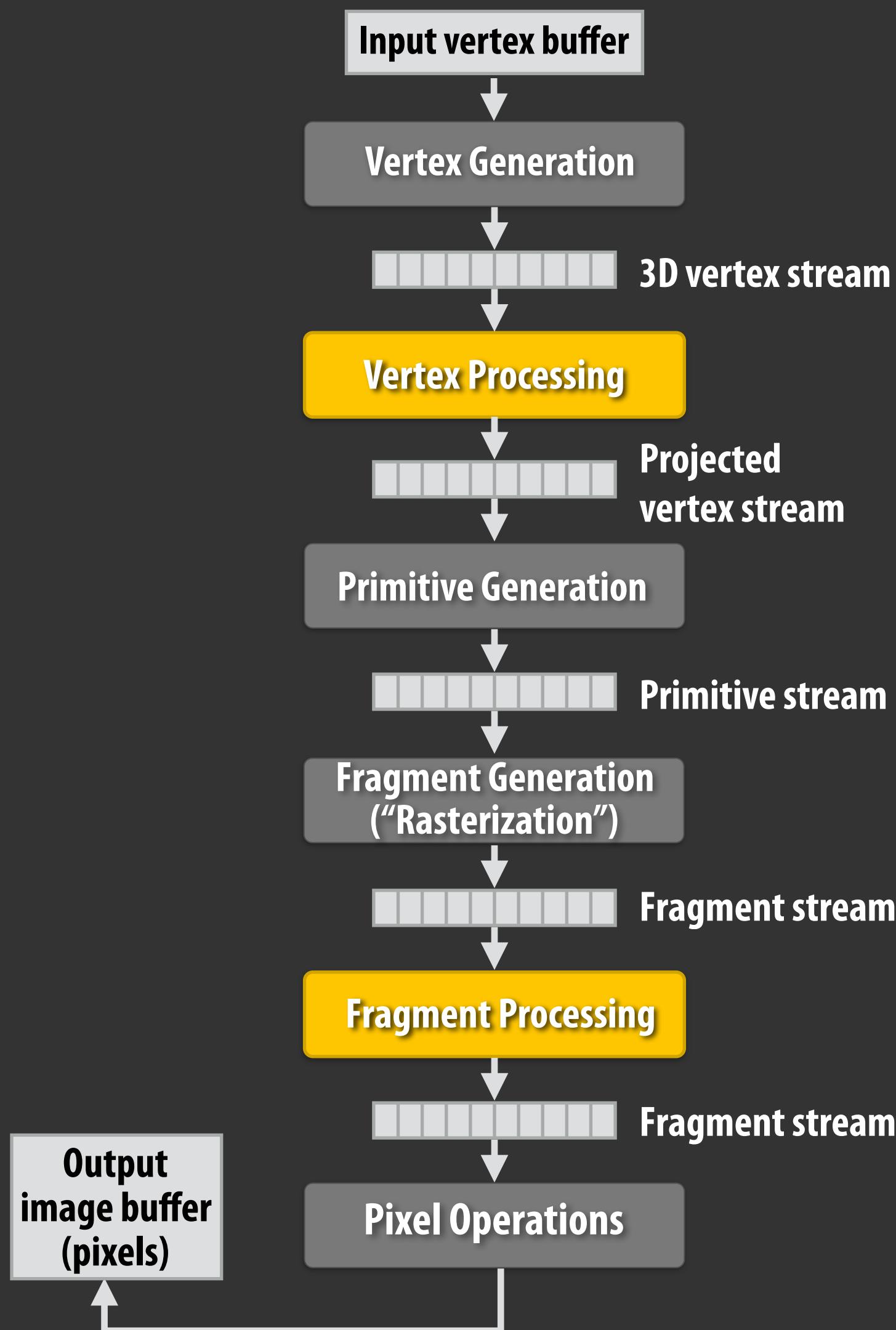
Custom circuits for key graphics arithmetic

Custom circuits for HW-assisted graphics-specific DRAM compression

HW logic for scheduling work onto these resources

Domain-specific languages for heterogeneous computing

OpenGL Graphics Pipeline (circa 2007)



The OpenGLTM Graphics System:
A Specification
(Version 1.0)

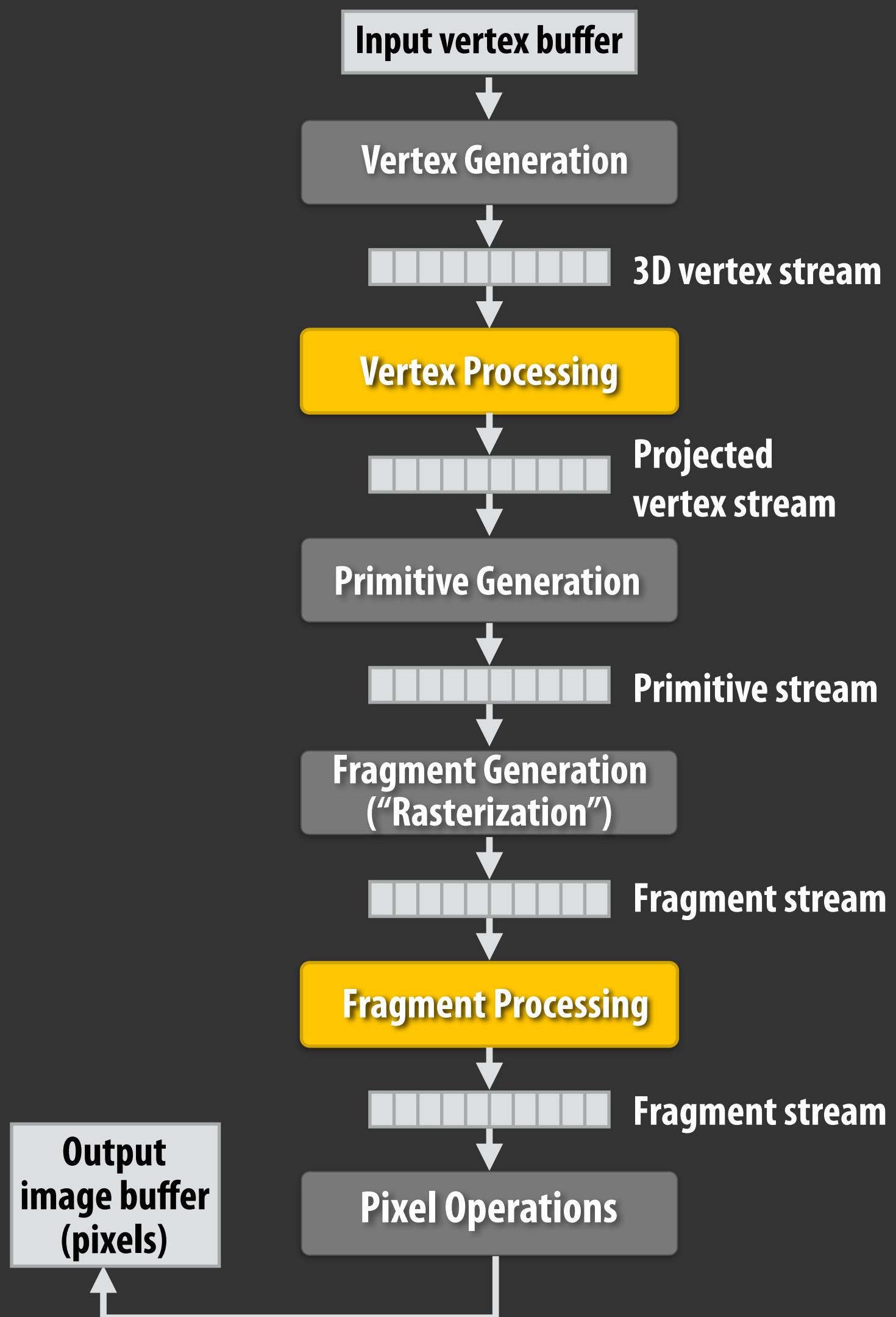
Mark Segal
Kurt Akeley

Editor:
Chris Frazier

Version 1.0 - 1 July 1994

Domain-specific languages for heterogeneous computing

OpenGL Graphics Pipeline (circa 2007)



```
uniform sampler2D myTexture;
uniform float3 lightDir;
varying vec3 norm;
varying vec2 uv;
```

```
void myFragmentShader()
{
    vec3 kd = texture2D(myTexture, uv);
    kd *= clamp(dot(lightDir, norm), 0.0, 1.0);
    return vec4(kd, 1.0);
}
```

"fragment shader"
(a.k.a kernel function mapped onto
input fragment stream)

read-only global variables

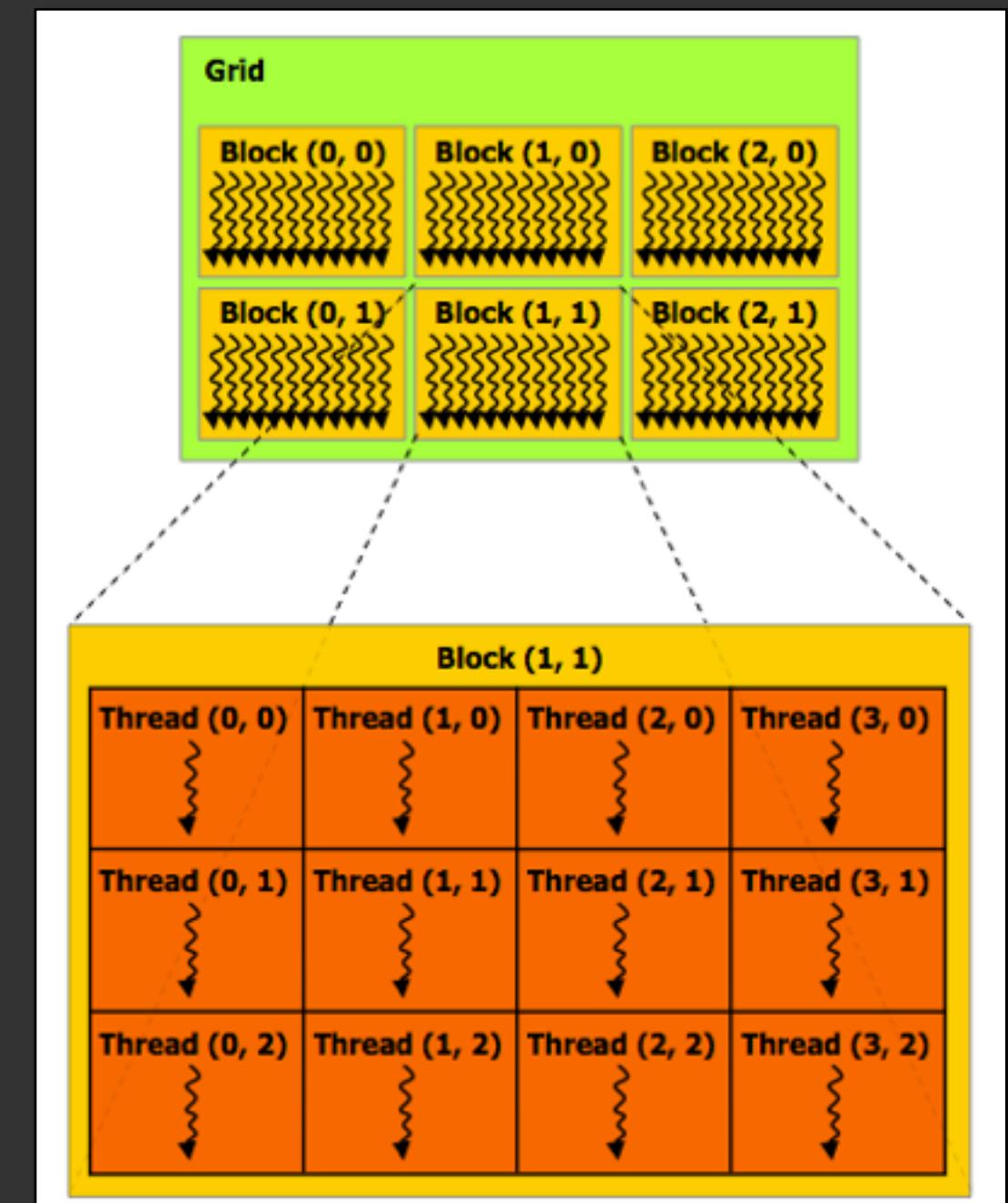
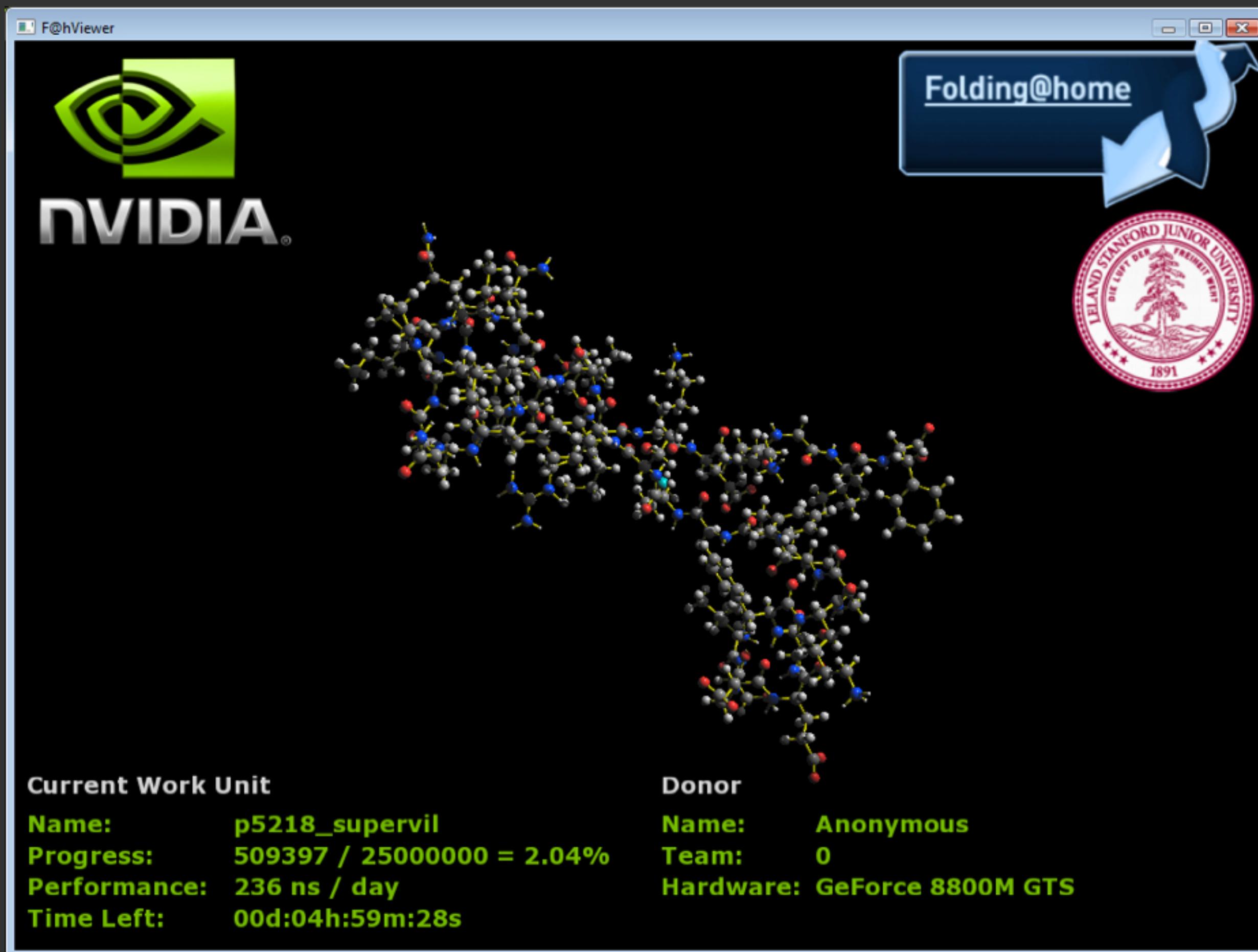
"per-element" inputs

per-element output:
RGBA surface color at pixel

Generalization beyond graphics: commodity parallel computing

Brook for GPUs (Buck 2004)

NVIDIA CUDA (2007)



Goals of visual computing (to date)

Modeling the real-world in increasingly rich detail: so we can simulate it (“render everything you’ve ever seen”)

Depict and organize information to augment human thought: enable humans to effectively use computing to create/analyze/interpret/communicate

Key characteristics of visual computing

Requires exceptional levels of efficiency

- Applications turn more ops/watt into new value
- Pack chips full of ALUs (parallel, heterogeneity/specialization are fundamental)
- Applications utilize hardware pipelines very well

Embrace domain-specific programming frameworks

- Achieve high efficiency/productivity
- Today: OpenGL, Halide, game engine frameworks, deep learning frameworks

Aspects of computation are fundamentally approximate

- Manifests as willingness to change algorithms (not approximate HW)

Visual computing — what's next?

Goals of visual computing (present — future)

To capture everything that can be seen

To enable humans to communicate more effectively

To record and analyze the world's visual information so
that computers can understand and reason about it

**The immediate future: capturing rich visual
information to enhance communication**

Capturing pixels to communicate

Ingesting/serving
the world's photos



2B photo uploads and shares
per day across Facebook sites
(incl. Instagram+WhatsApp)
[FB2015]

Ingesting/streaming
world's video



PSY - GANGNAM STYLE (강남스타일) M/V

Youtube 2015: 300 hours
uploaded per minute [Youtube]

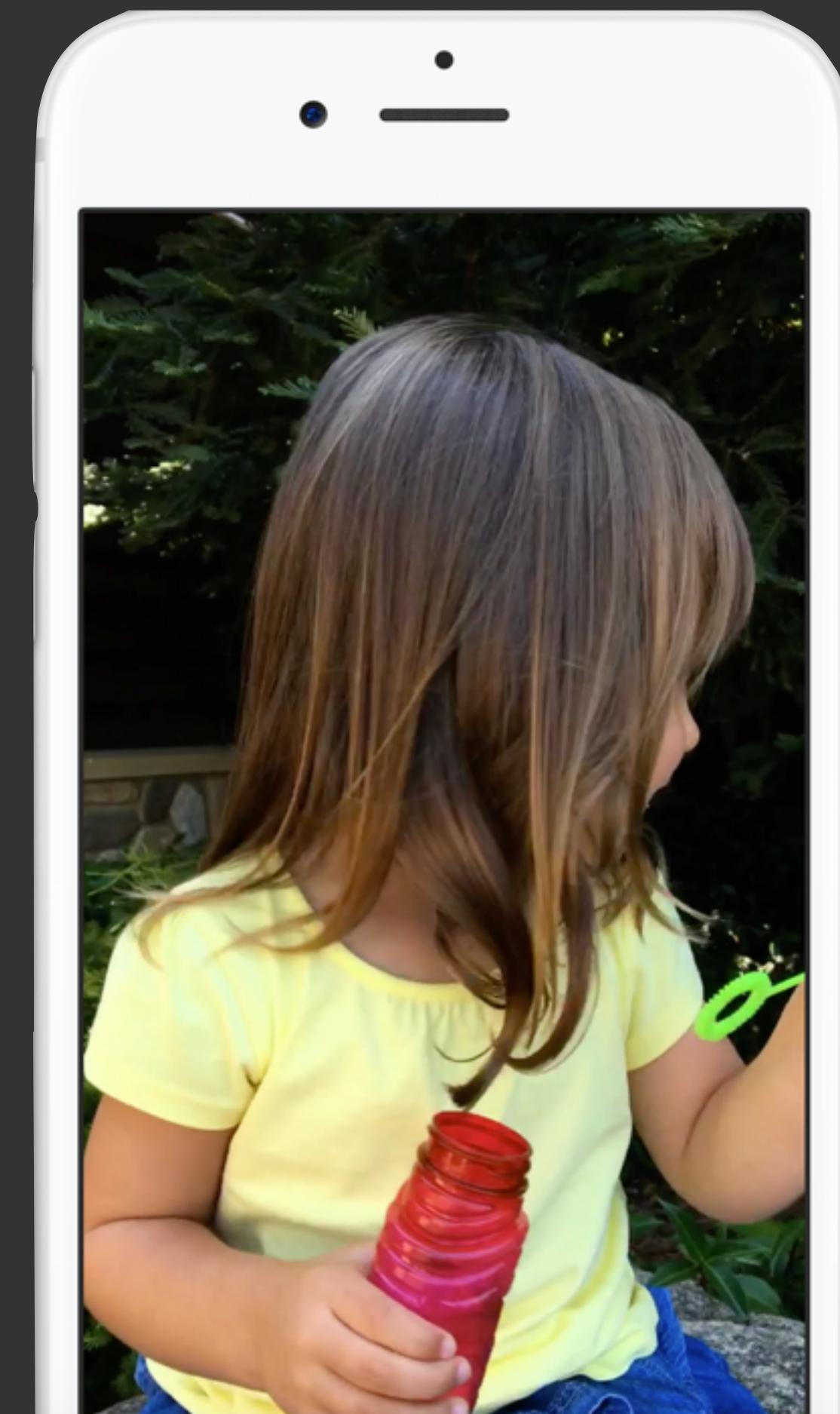
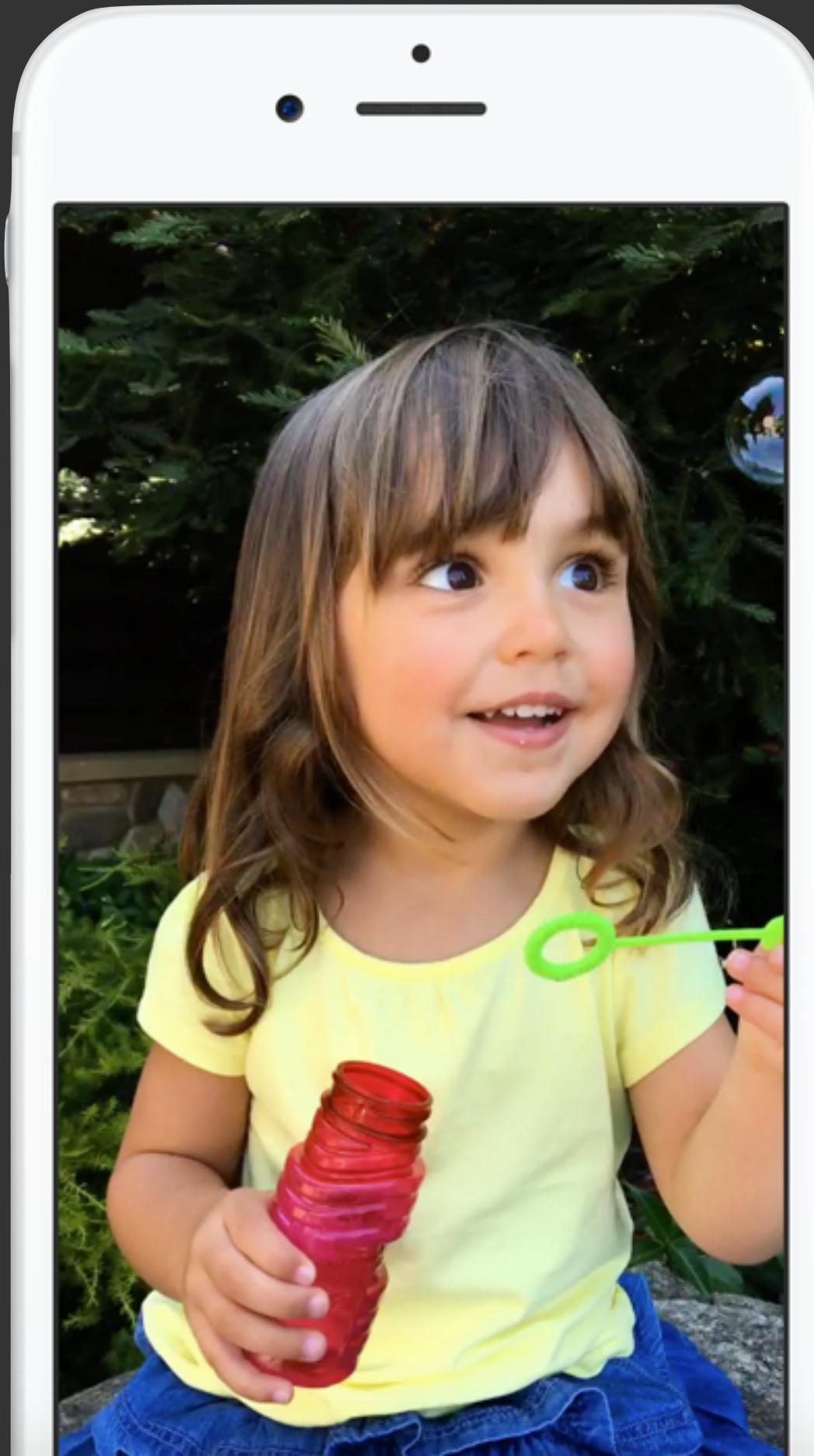
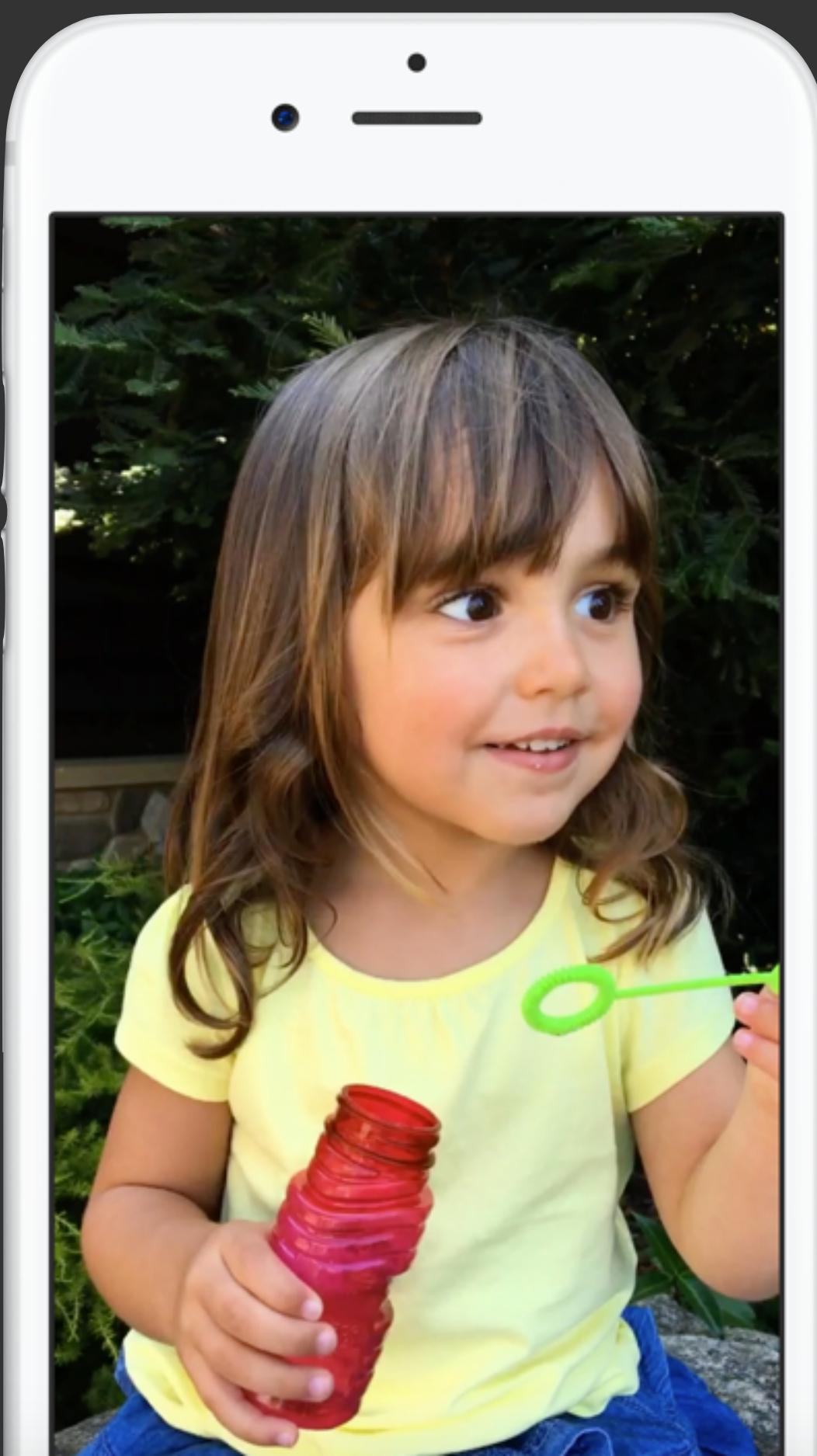
Cisco VNI projection:
80-90% of 2019 internet
traffic will be video.
(64% in 2014)

The collage includes:

- Twitch:** Shows the "Top Live Channels" page with multiple game streams like League of Legends and Dota 2.
- YouTube:** A video player showing PSY's 'Gangnam Style' M/V with a timestamp of 0:35 / 4:12.
- Vine:** A grid of short video loops, including a baseball game, a soccer goal, and a golfer.

Richer content: beyond a single image

- Example: Apple's "Live Photos"
- Each photo is not only a single frame, but a few seconds of video before and after the shutter is clicked



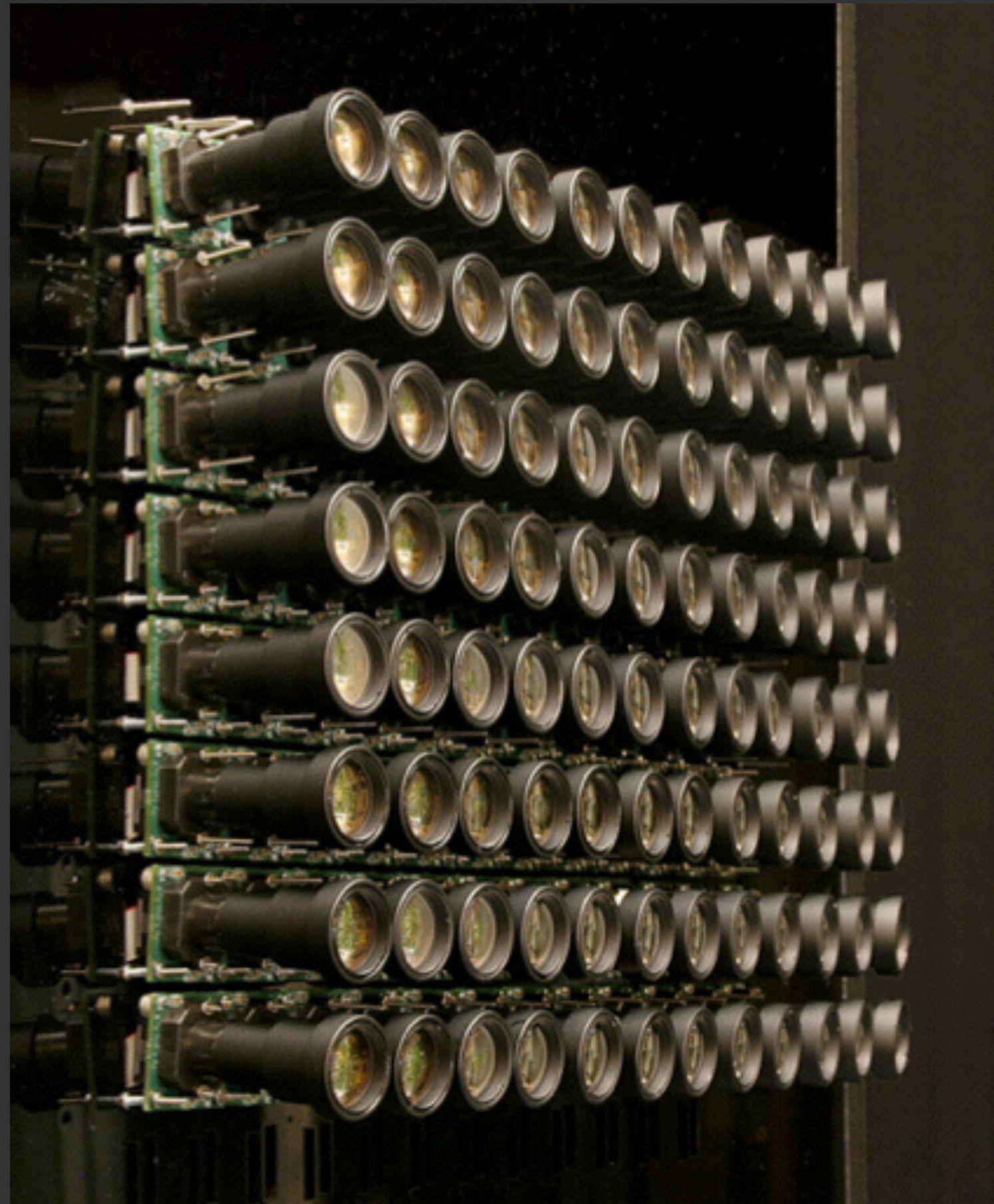
Facebook Live



Acquiring richer content: light fields



Stanford camera array
Wilburn [2005]



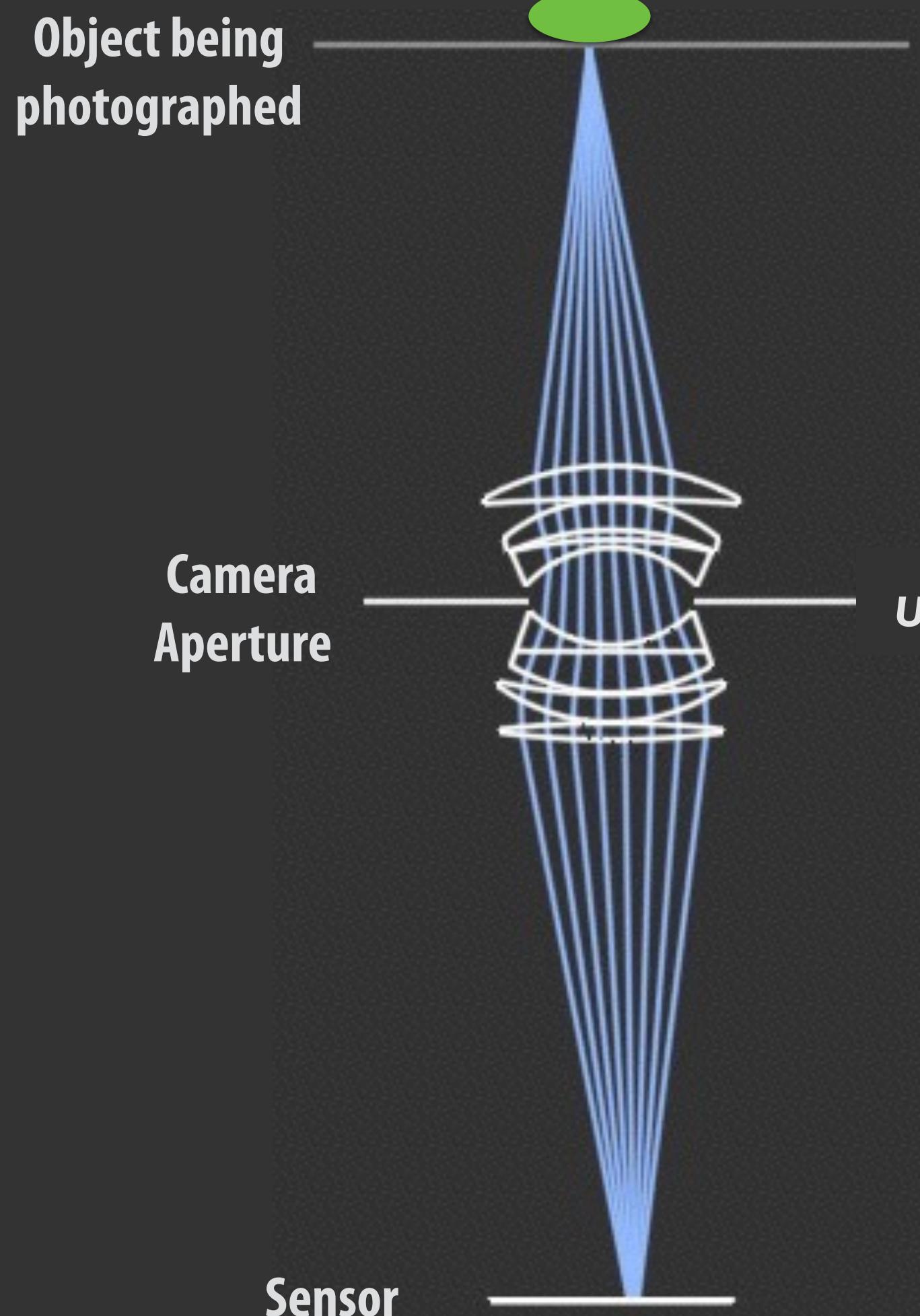
Richer content: light fields



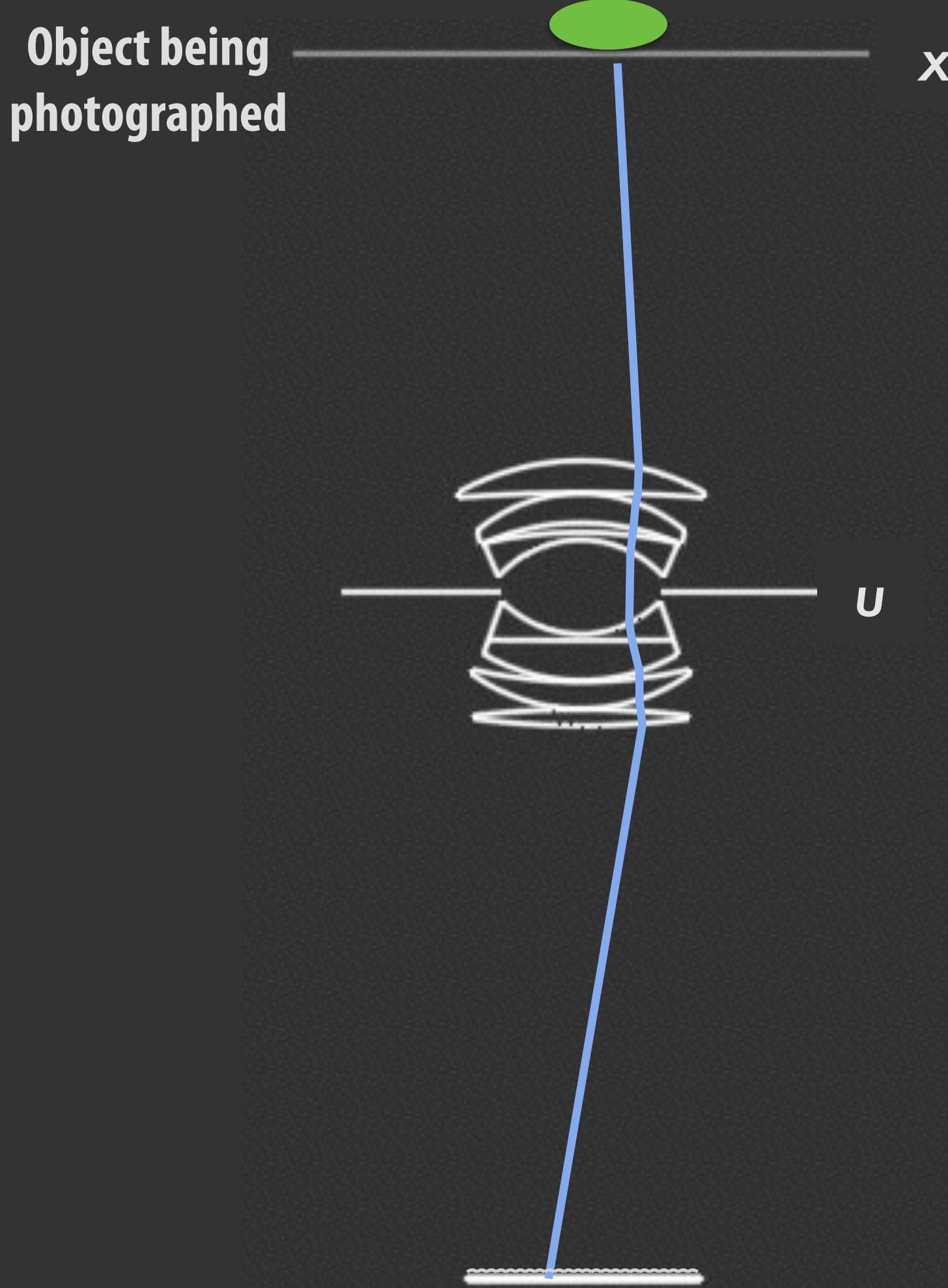
Light L16

Lytro Illum

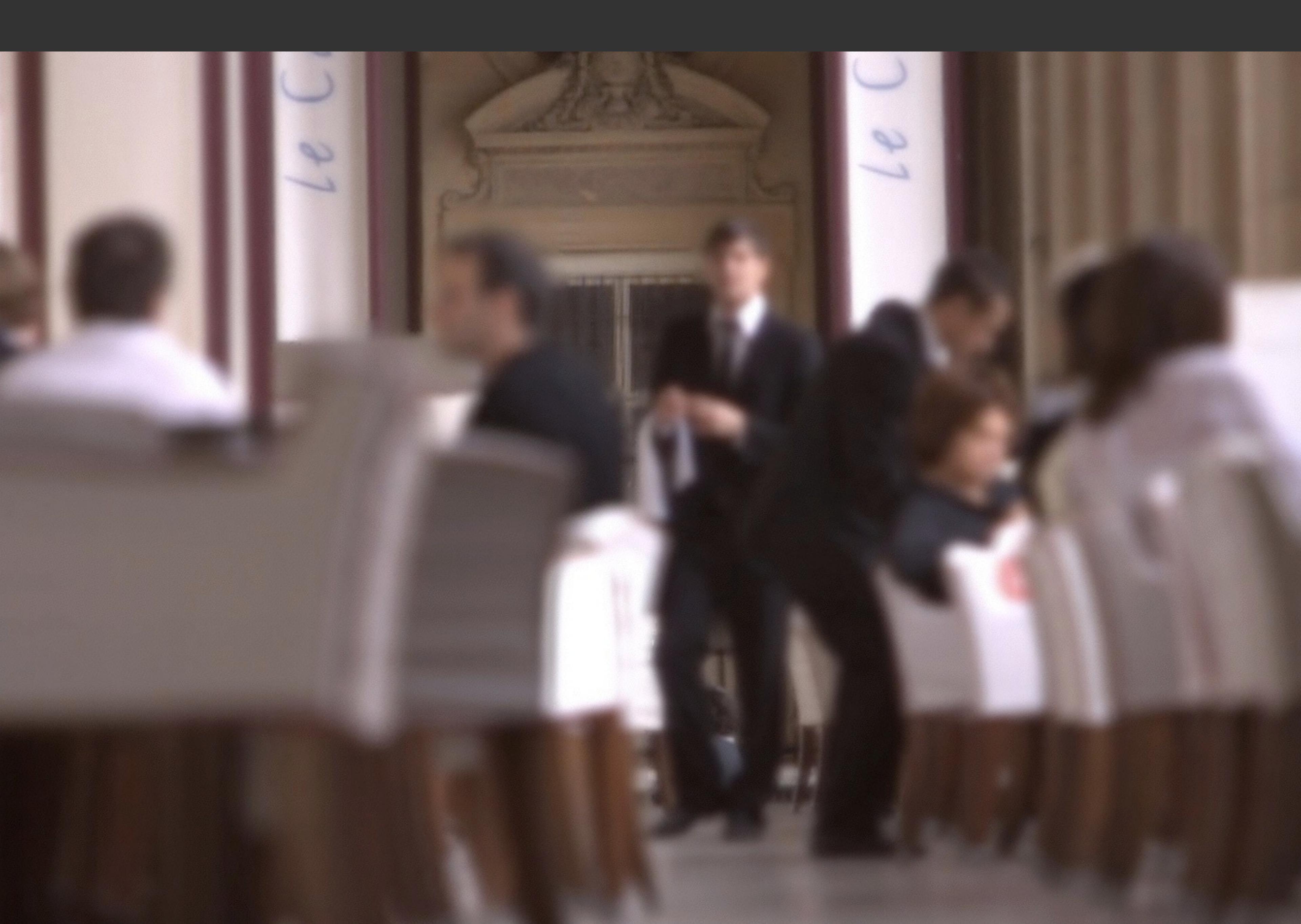
Light field camera: capturing a light field



2D traditional camera:
measures how much light hits a
point on sensor



"4D" light field camera:
measures how much light hits point
on sensor **from a particular direction**



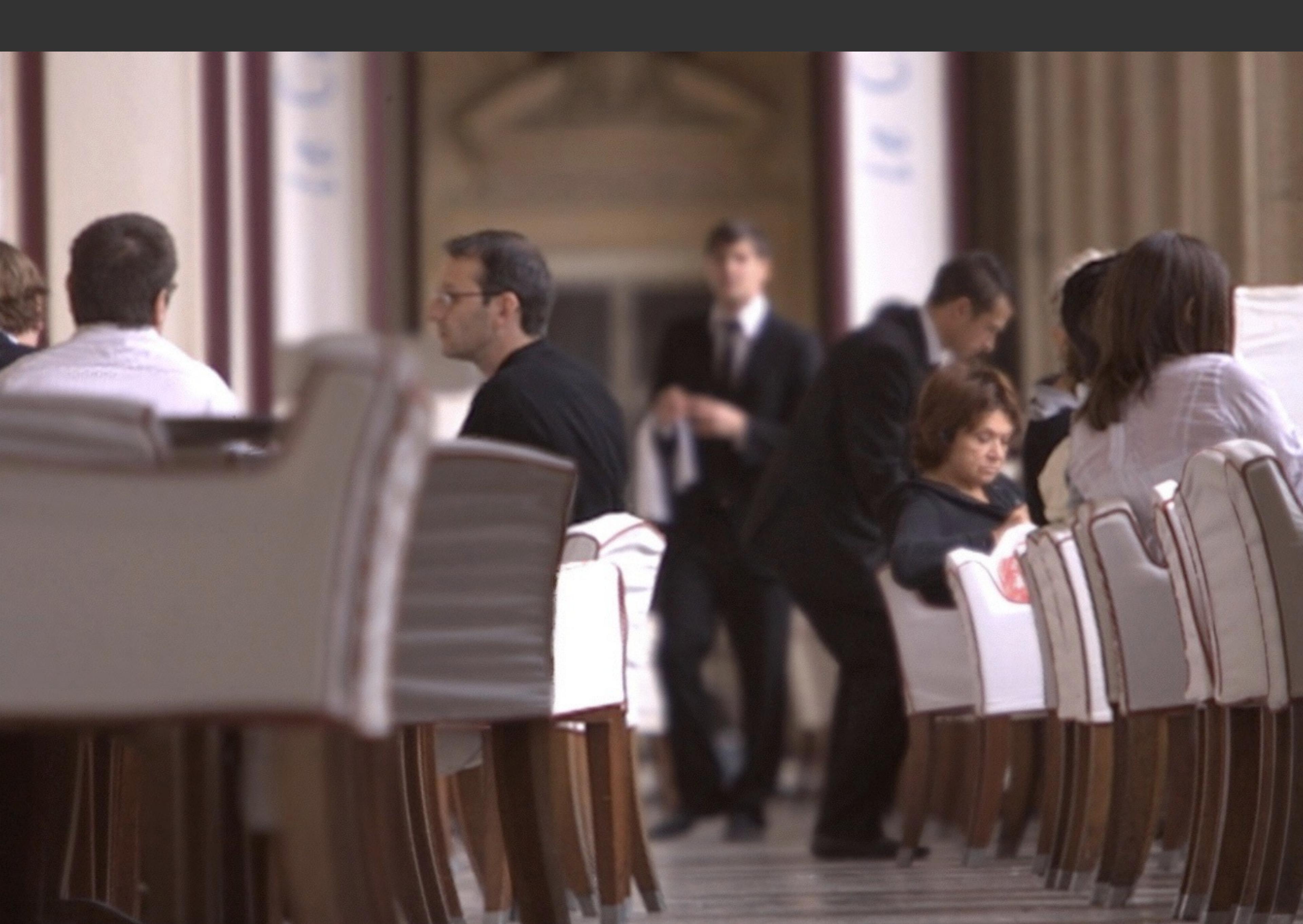
[Slide courtesy Ren Ng]



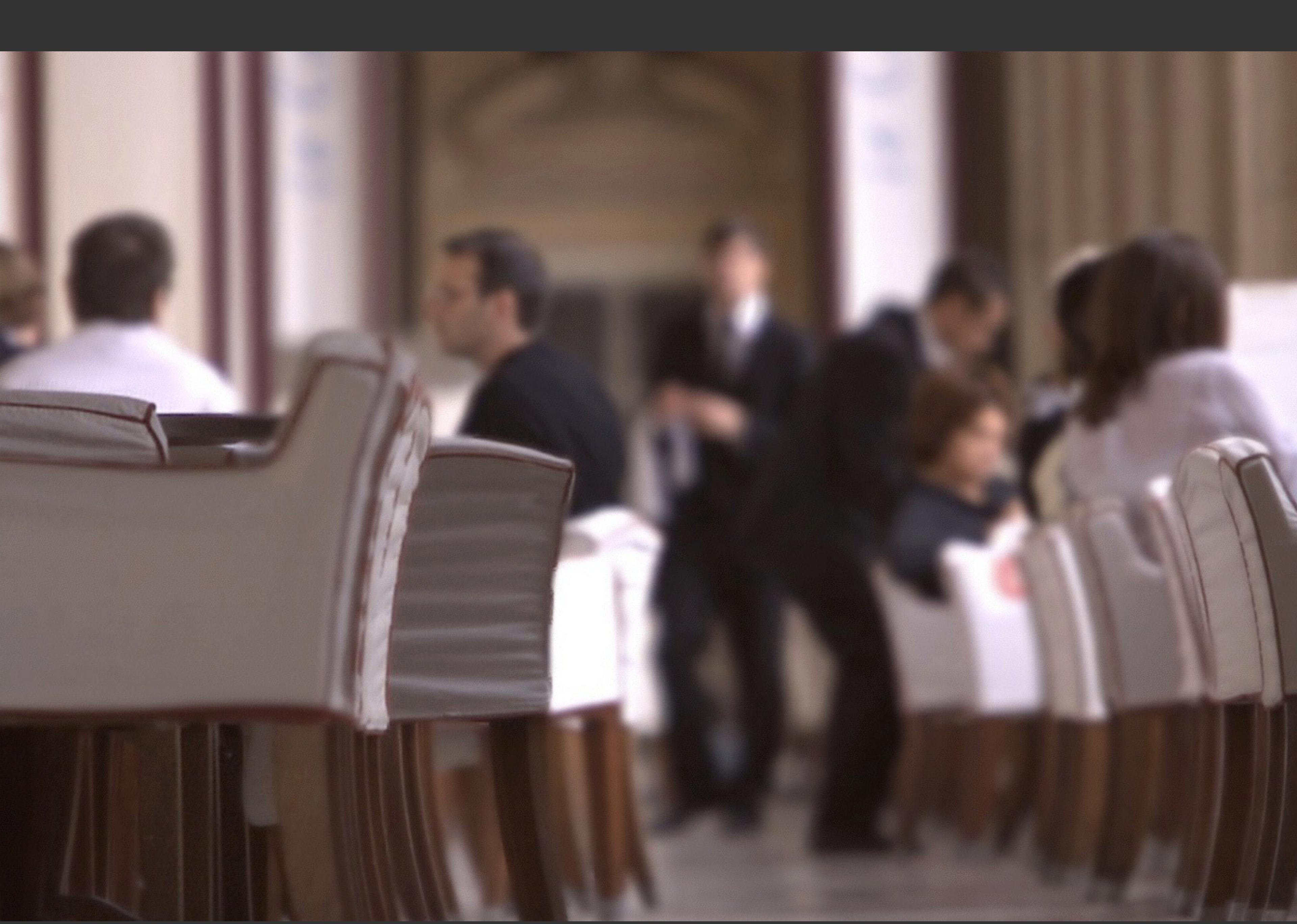
[Slide courtesy Ren Ng]



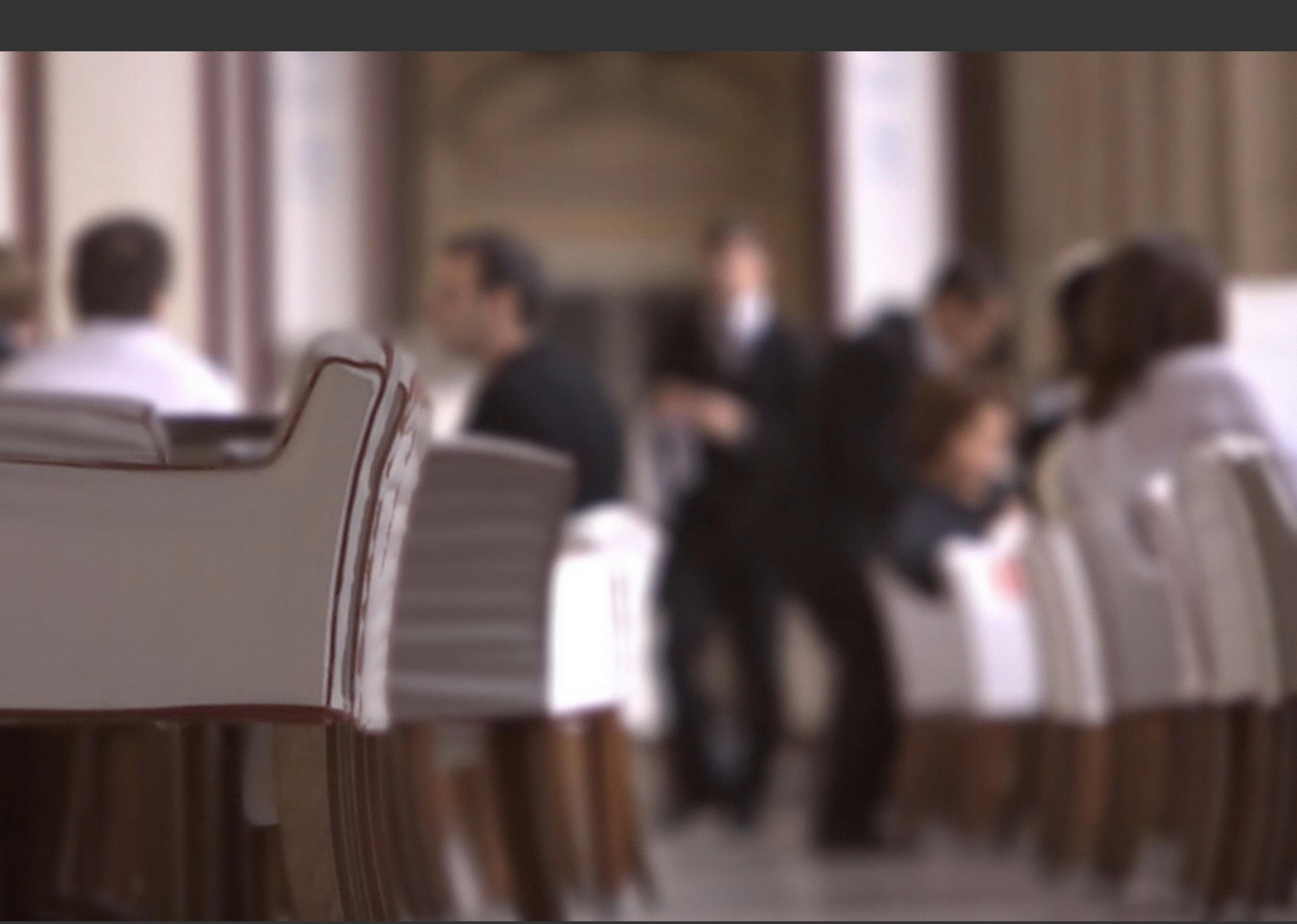
[Slide courtesy Ren Ng]



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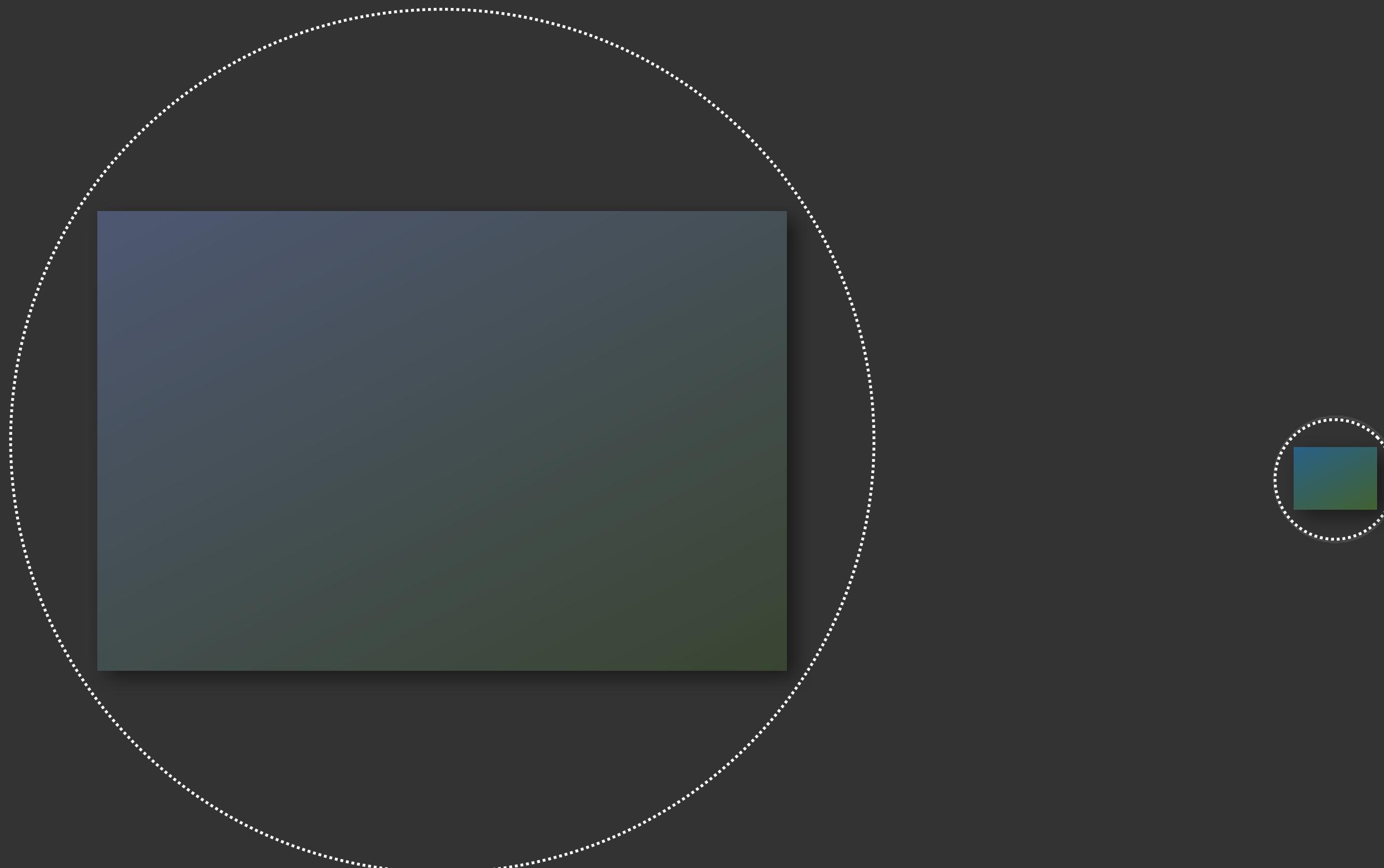


[Slide courtesy Ren Ng]



[Slide courtesy Ren Ng]

Sensor industry has large untapped resolution



Full-Frame Sensor

36 x 24 mm

Up to 36 MP

4.9 micron pixel

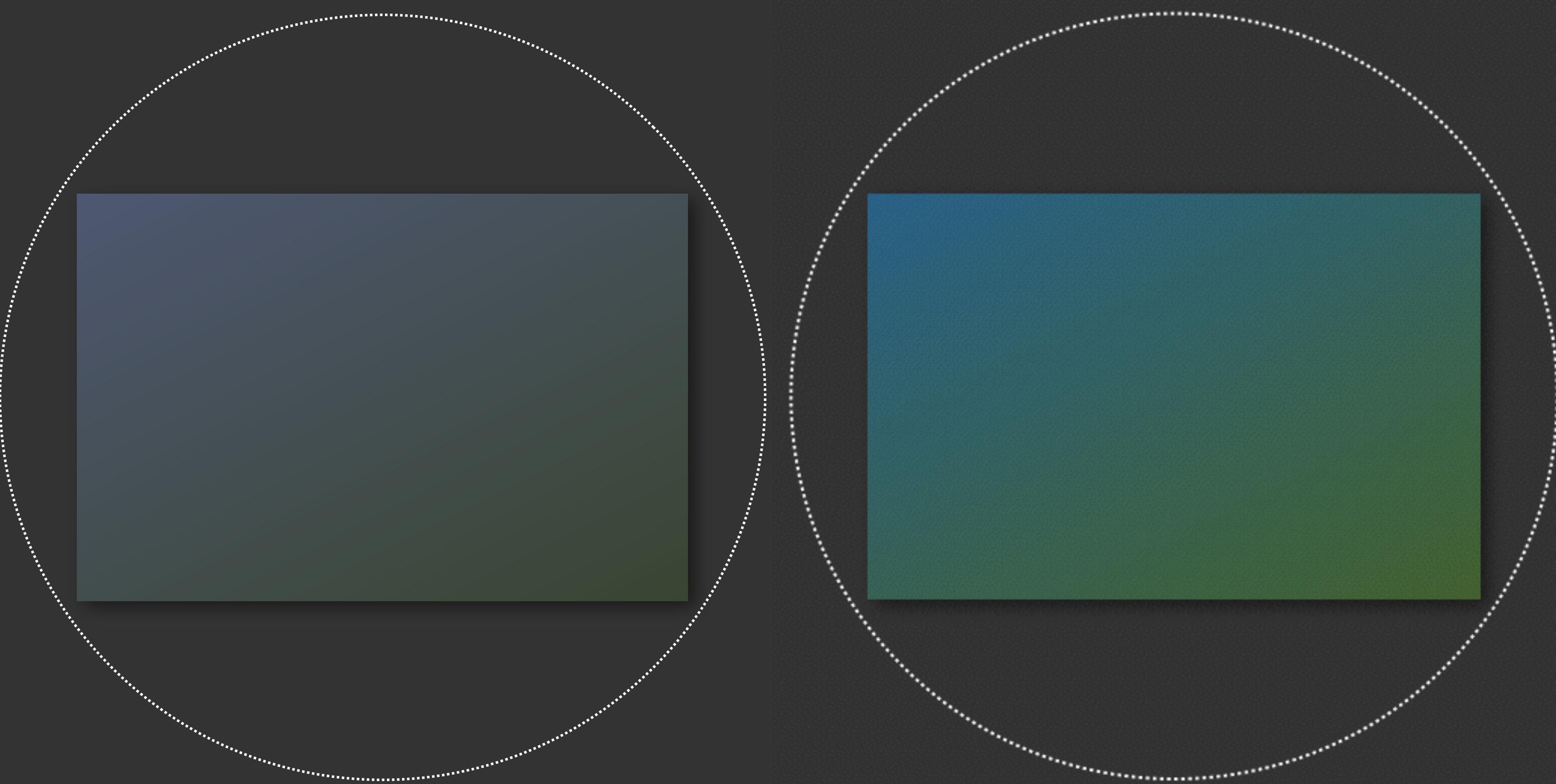
1/3" Sensor

4.8 x 3.6 mm

Up to 13 MP

1.12 micron pixel

Sensor industry has large untapped resolution



Full-Frame Sensor

36 x 24 mm

Up to 36 MP

4.9 micron pixel

Full-Frame Sensor

36 x 24 mm

688 MP

1.12 micron pixel

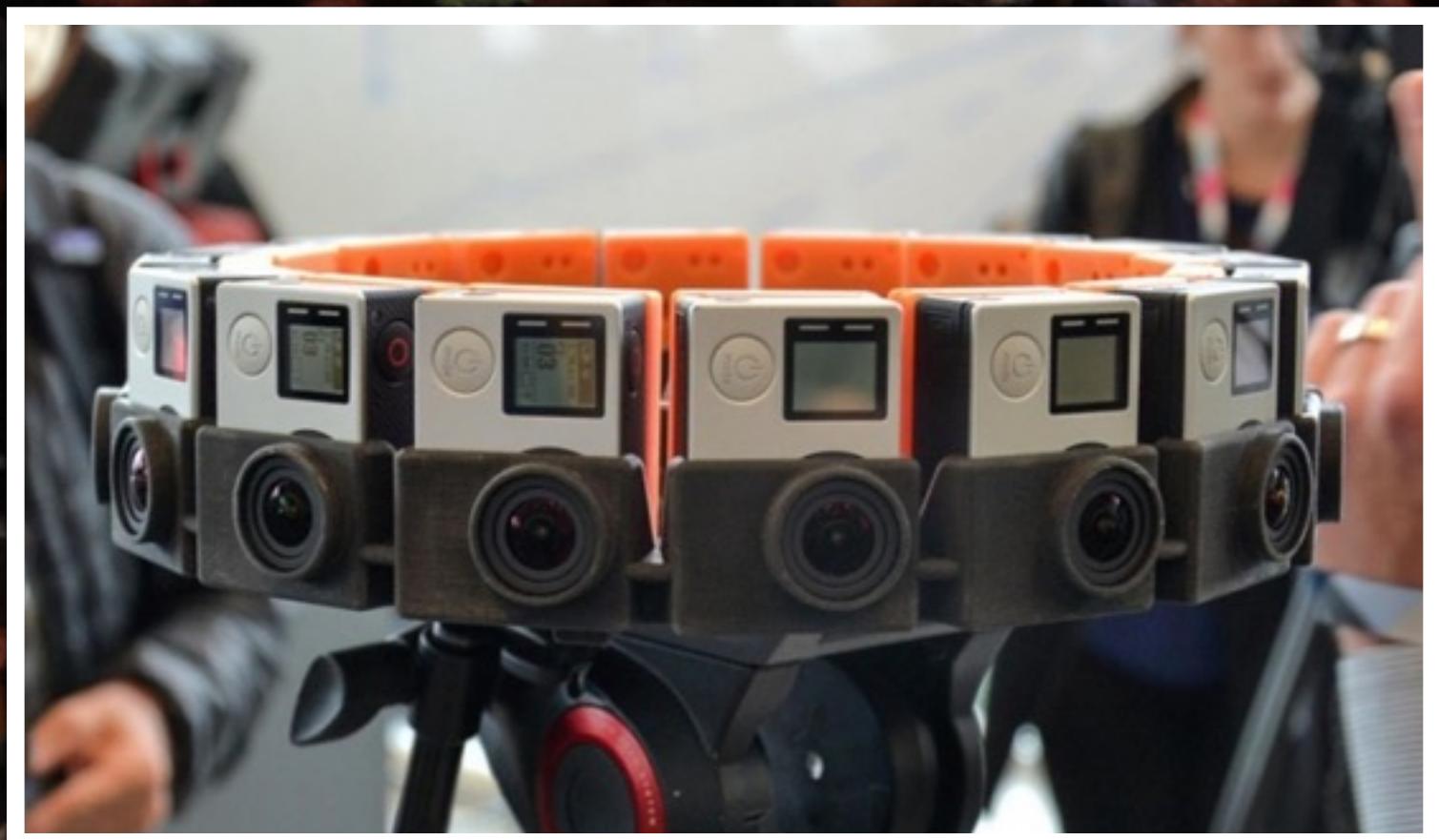
Lytro Cinema

755 Mpixel camera



VR output

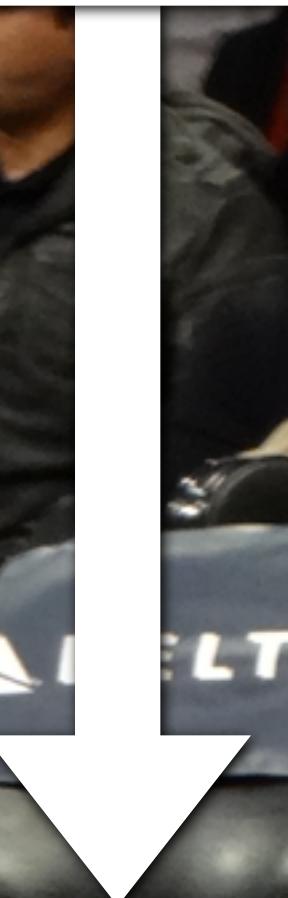




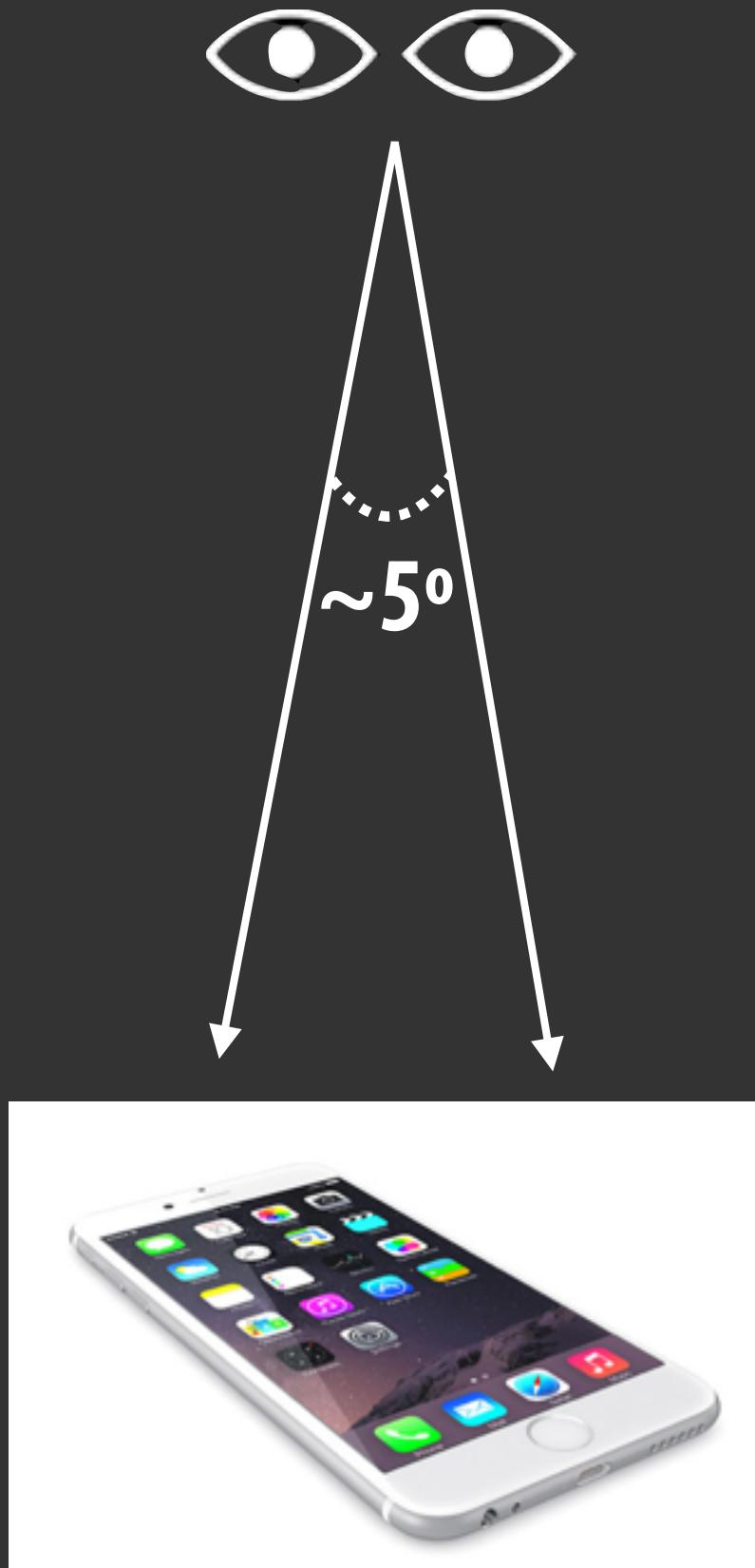
Example: Google's JumpVR video

Input stream: 16 4K GoPro cameras

Register + 3D align video stream (on edge device)
Broadcast encoded video stream across
the country to millions of viewers



VR creates high resolution requirements



Future “retina” VR display:
57 ppd covering 180°
= $10K \times 10K$ display per eye
= 200 MPixel

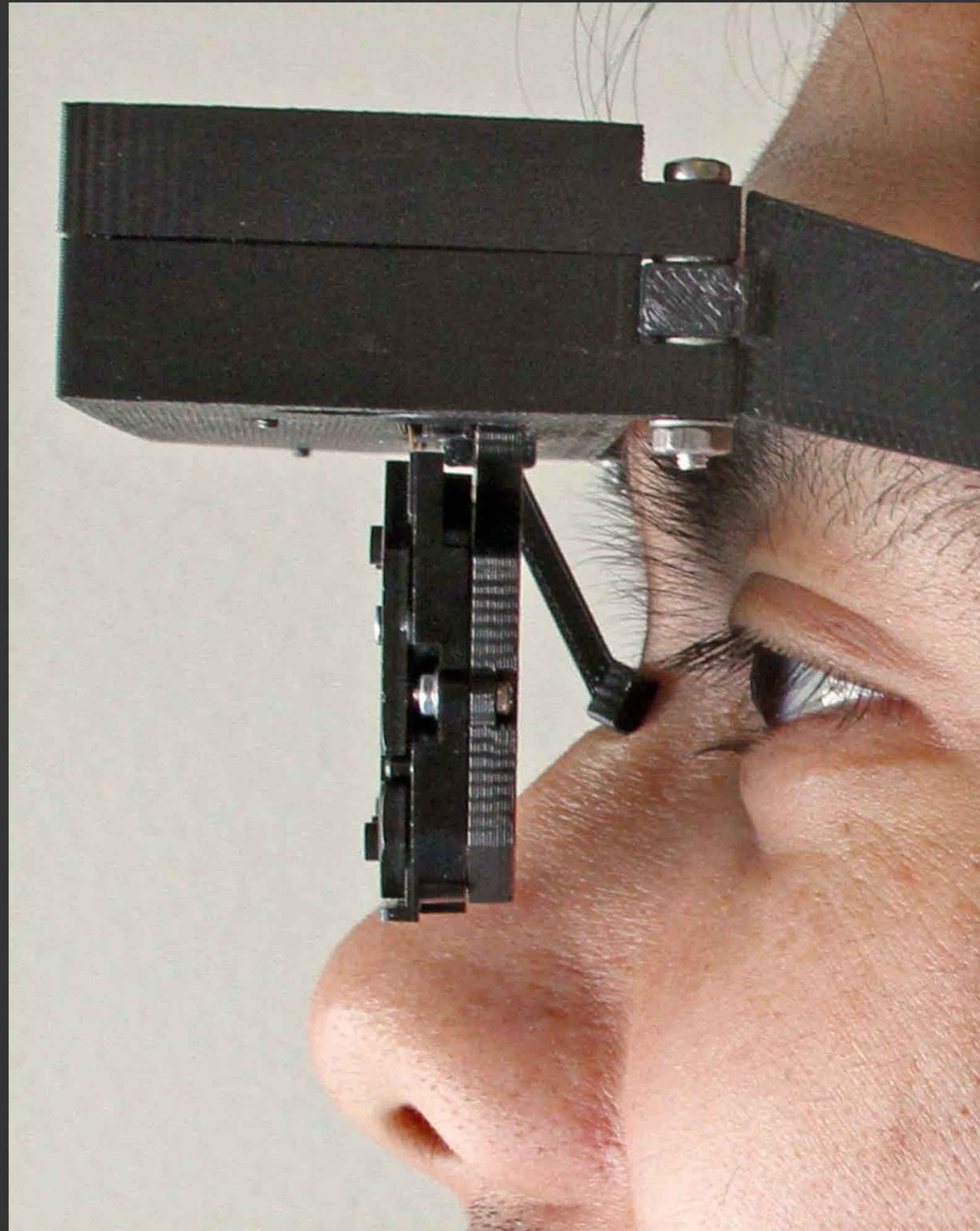
RAW data rate @ 120Hz ≈ 72 GB/sec

iPhone 6: 4.7 in “retina” display:

1.3 MPixel

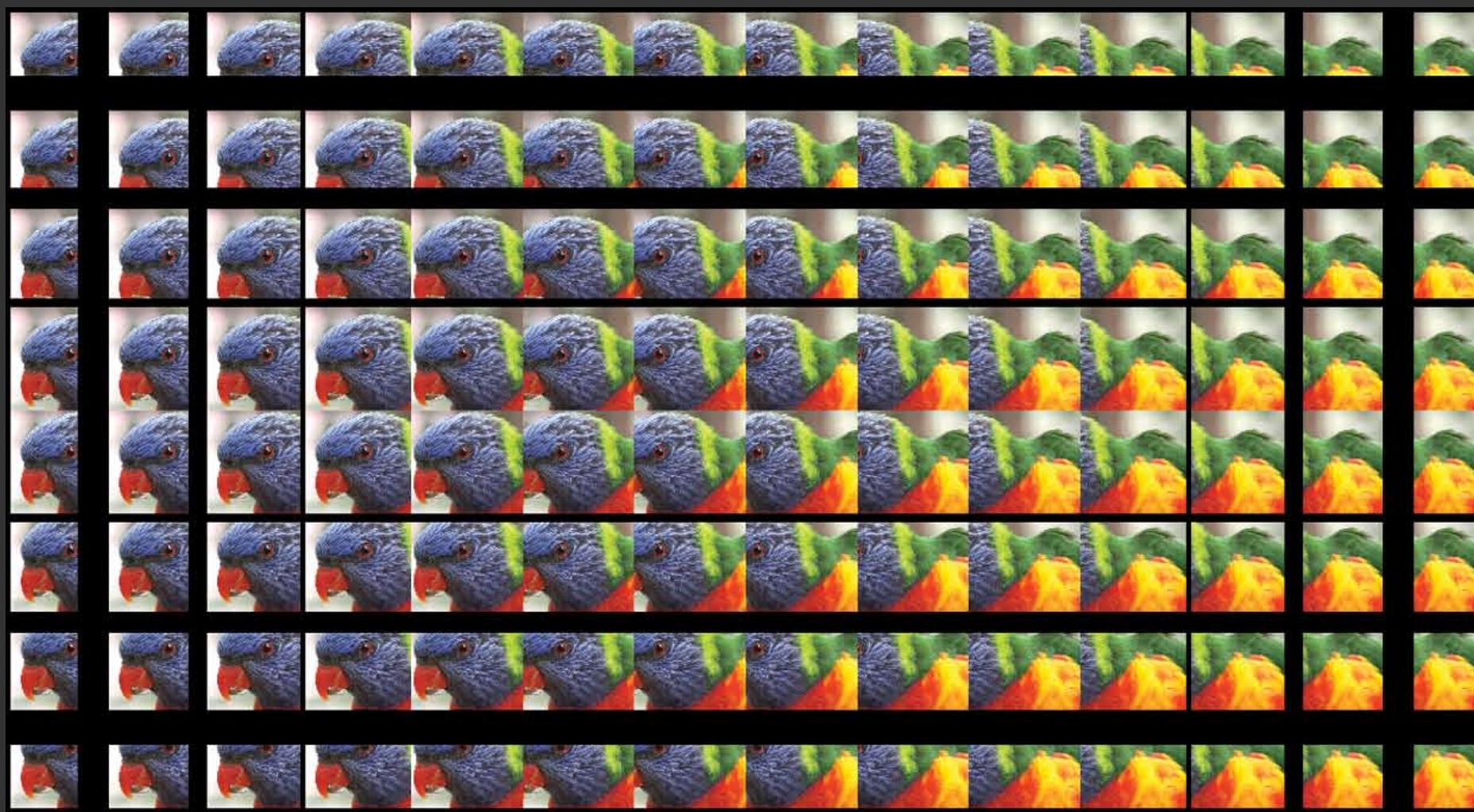
326 ppi \rightarrow 57 ppd

VR: Light field display



146 x 78 spatial resolution
Using 1MP microdisplay

Simple idea:
Recreate the same light field that was present in the scene when it was captured



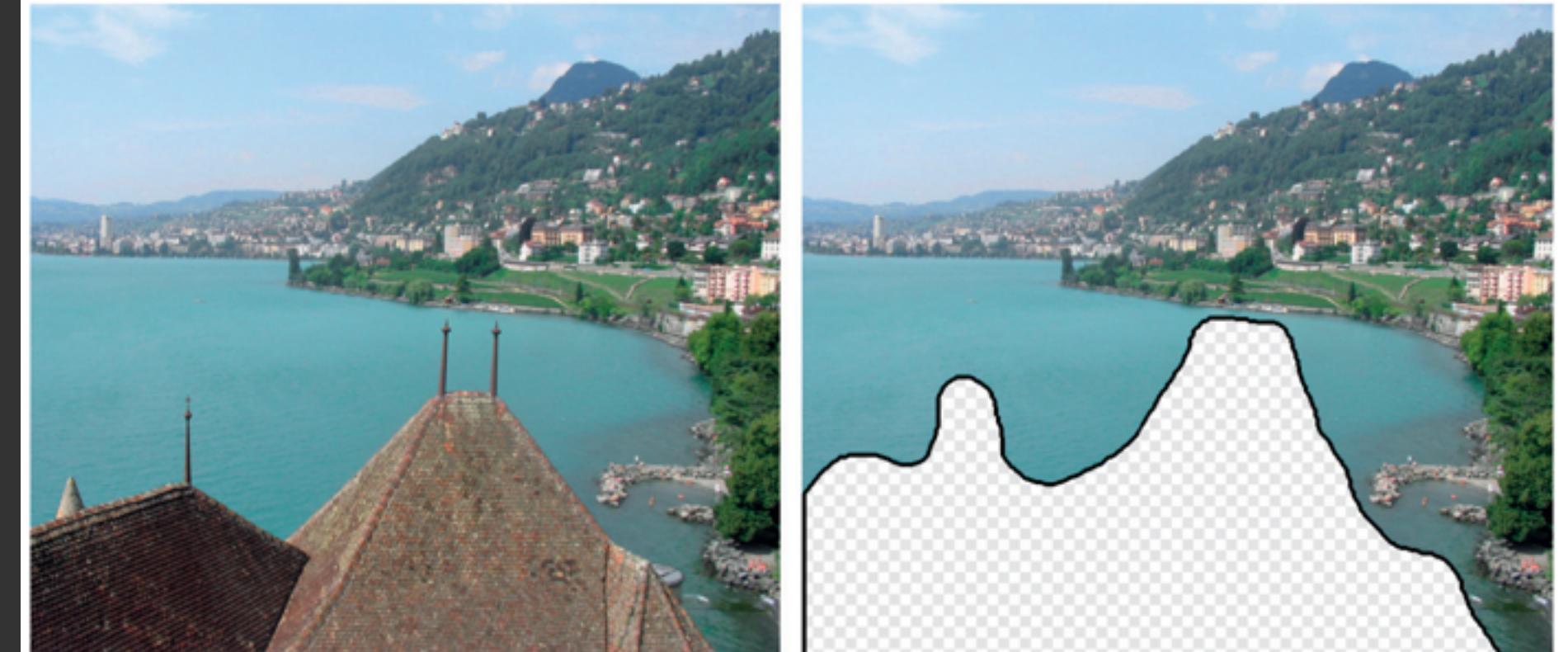
Output of display (prior to optics)

Enhancing communication: understanding images to improve acquired content

AutoEnhance:

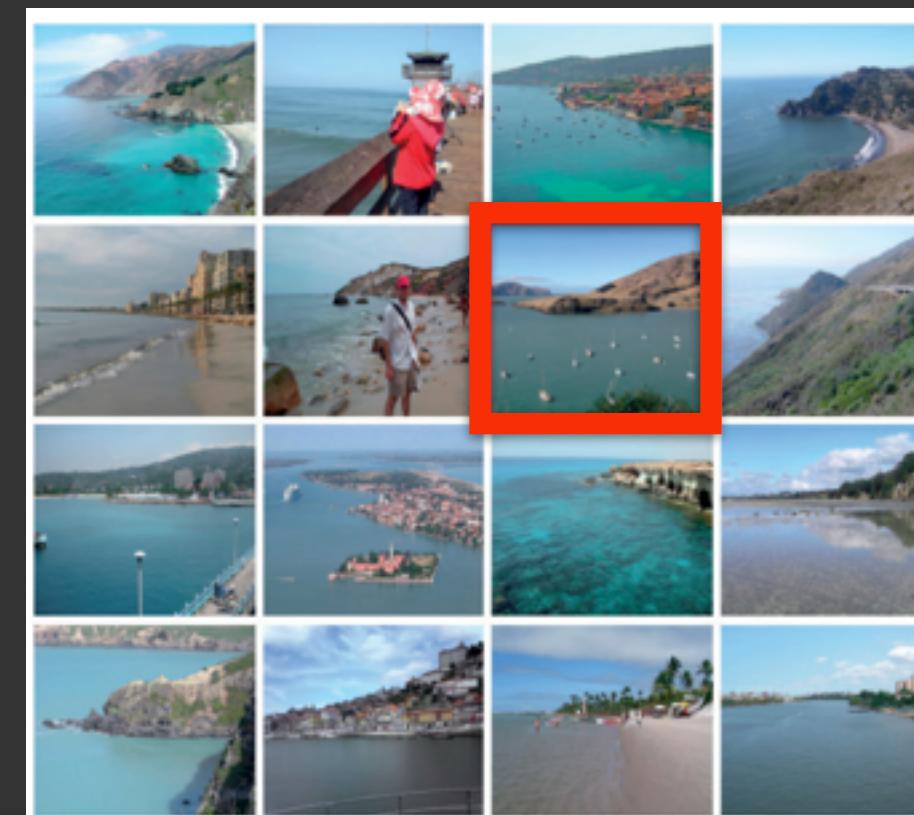


Photo “fix up” [Hayes 2007]



My bad vacation photo

Part to fix



Similar photos others
have taken



Fixed!

Summary

We are observing rapid growth in the richness of visual communication

Sensing the world with higher fidelity to deliver improved content to humans

2030 challenge: recording and analyzing the world's visual information, so computers can understand and reason about it

Capturing everything about the visual world

To understand people

To understand the world around vehicles/drones

To understand cities

Mobile

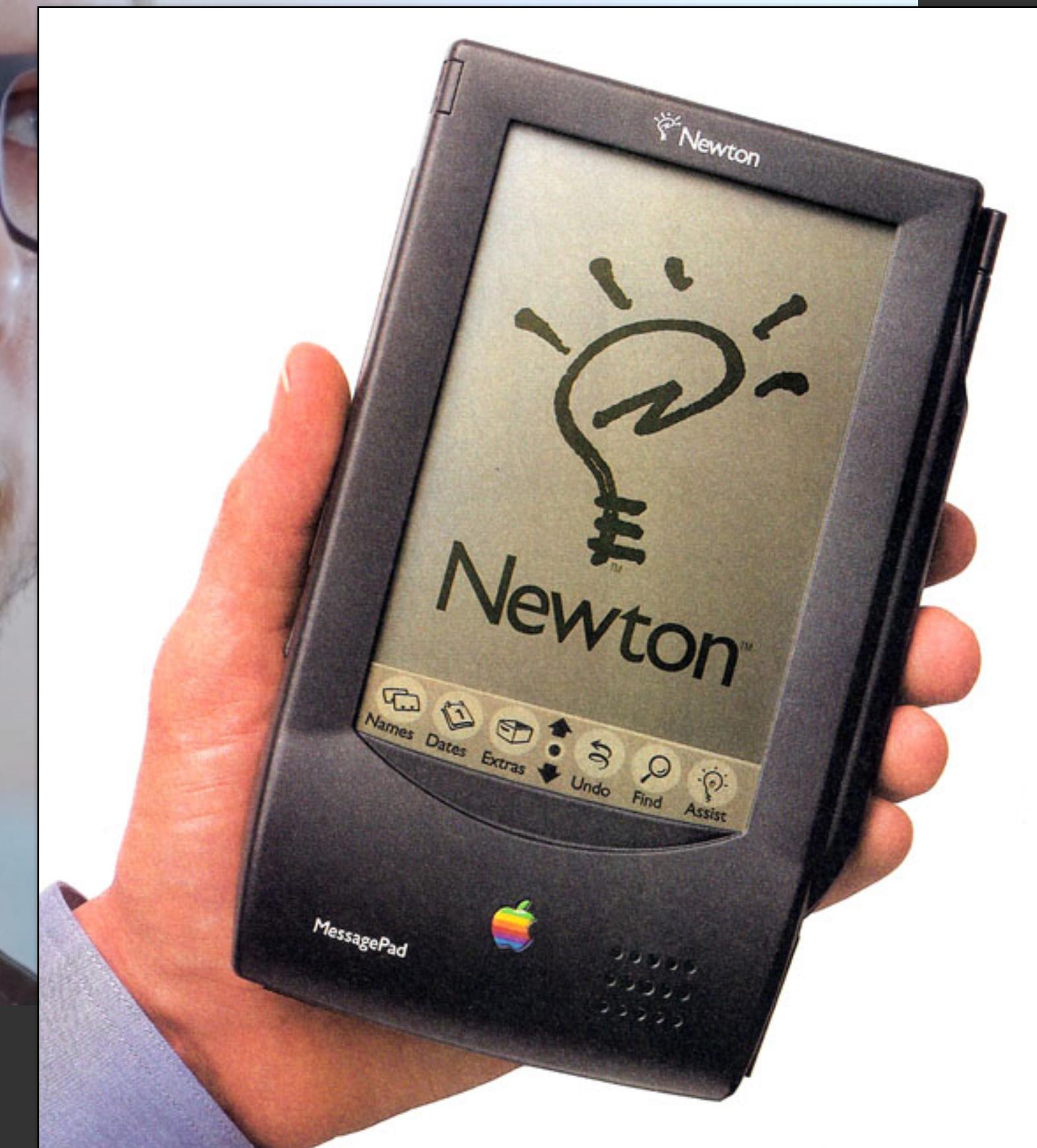
Continuous (always on)

Exceptionally high resolution

Capture for computers to analyze, not humans to watch

Capturing images to understand humans
**(why there will be high-resolution camera(s) always on,
on every human)**

Google Glass



Gwangjang Market (Seoul)



What does this say?



What is this?



Is it okay for me to sit there?

Is this woman annoyed that I sat down beside her? (Am I offending anyone?)

Why is she staring at me?

Should I attempt to greet the individuals at my table? (are they in a conversation that should not be interrupted)

When is a socially appropriate time to interrupt?



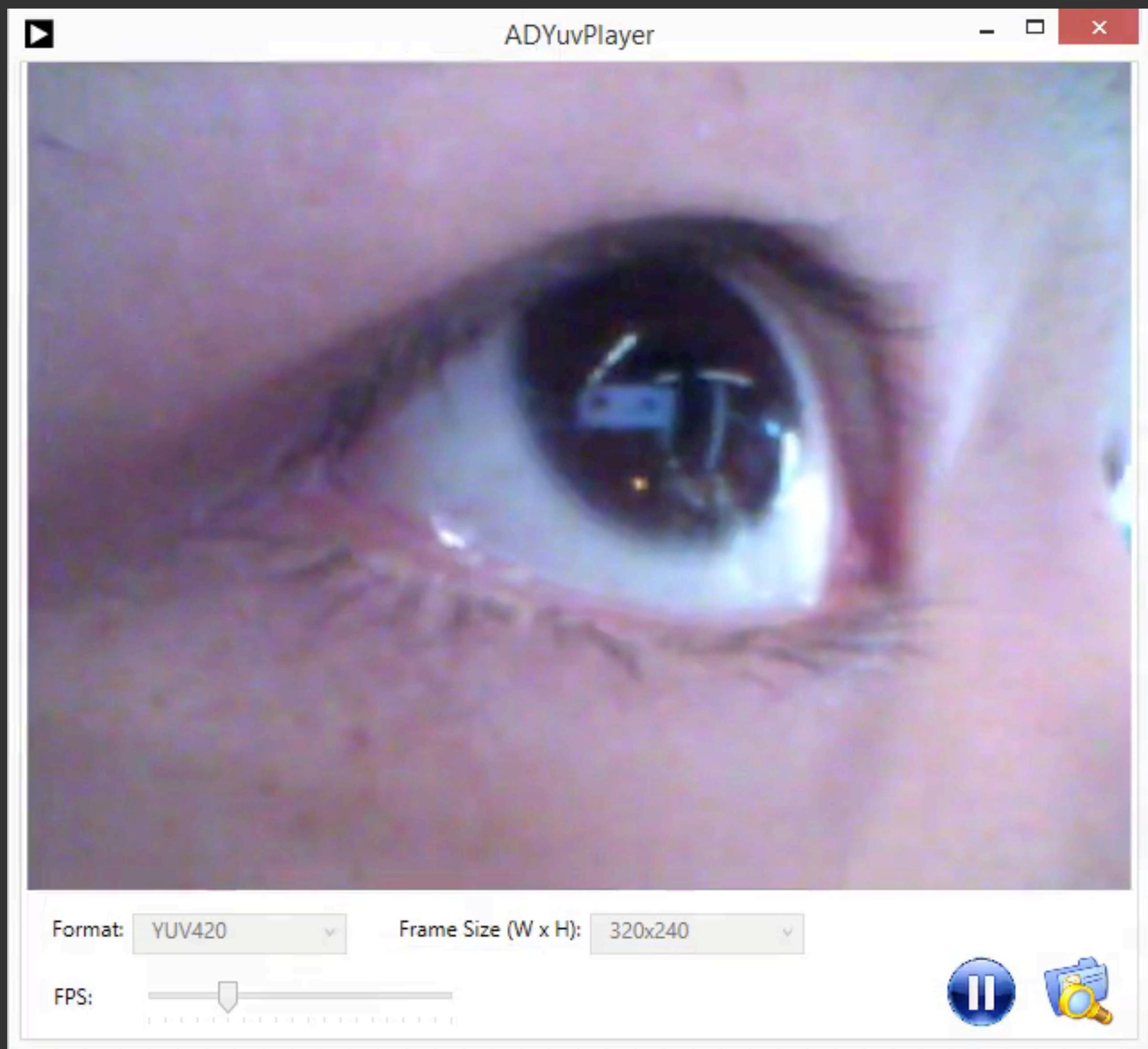
A future digital assistant must capture and comprehend extremely subtle aspects of human social behavior

Body language

Eye movement

Social context

Capturing / tracking eye movement



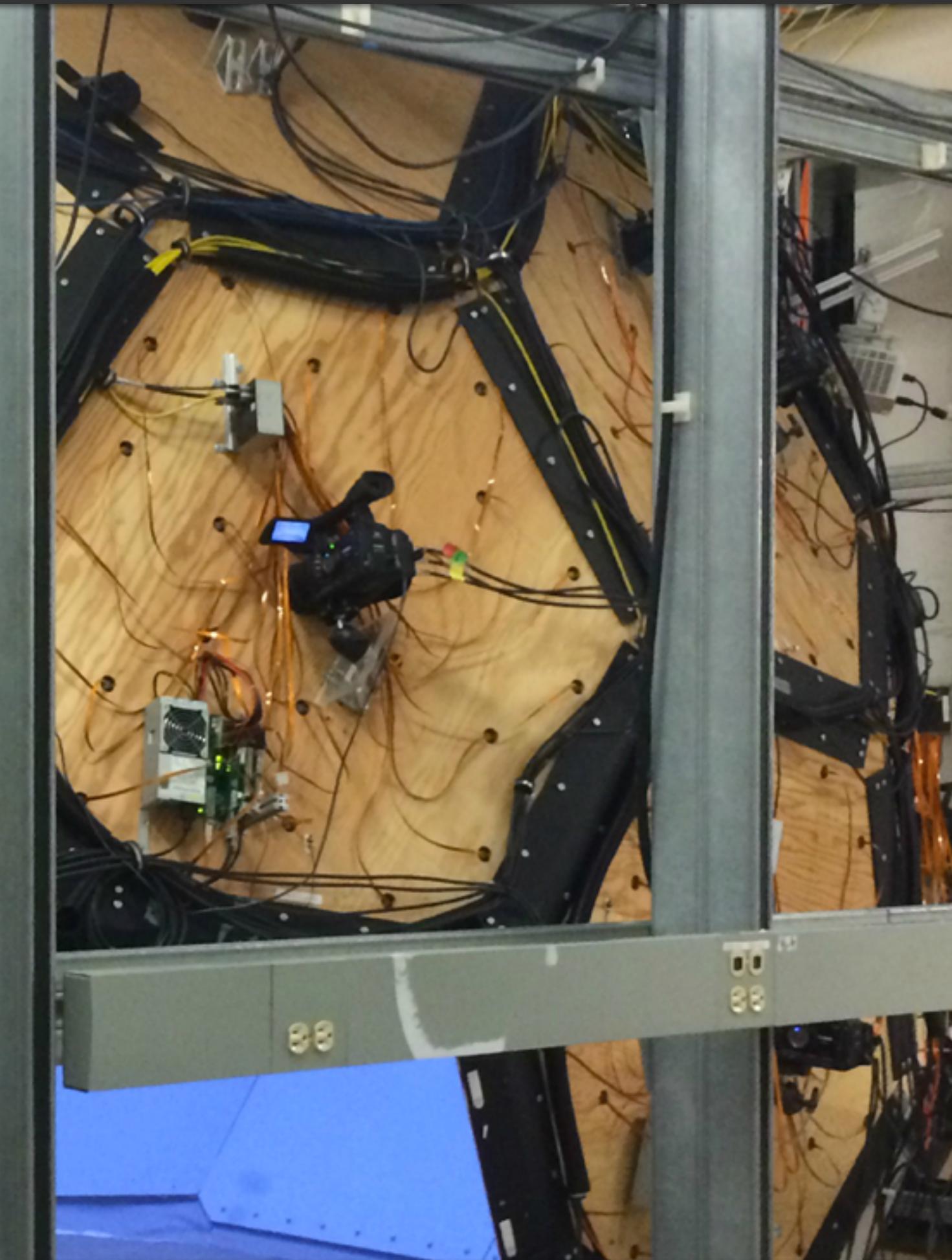
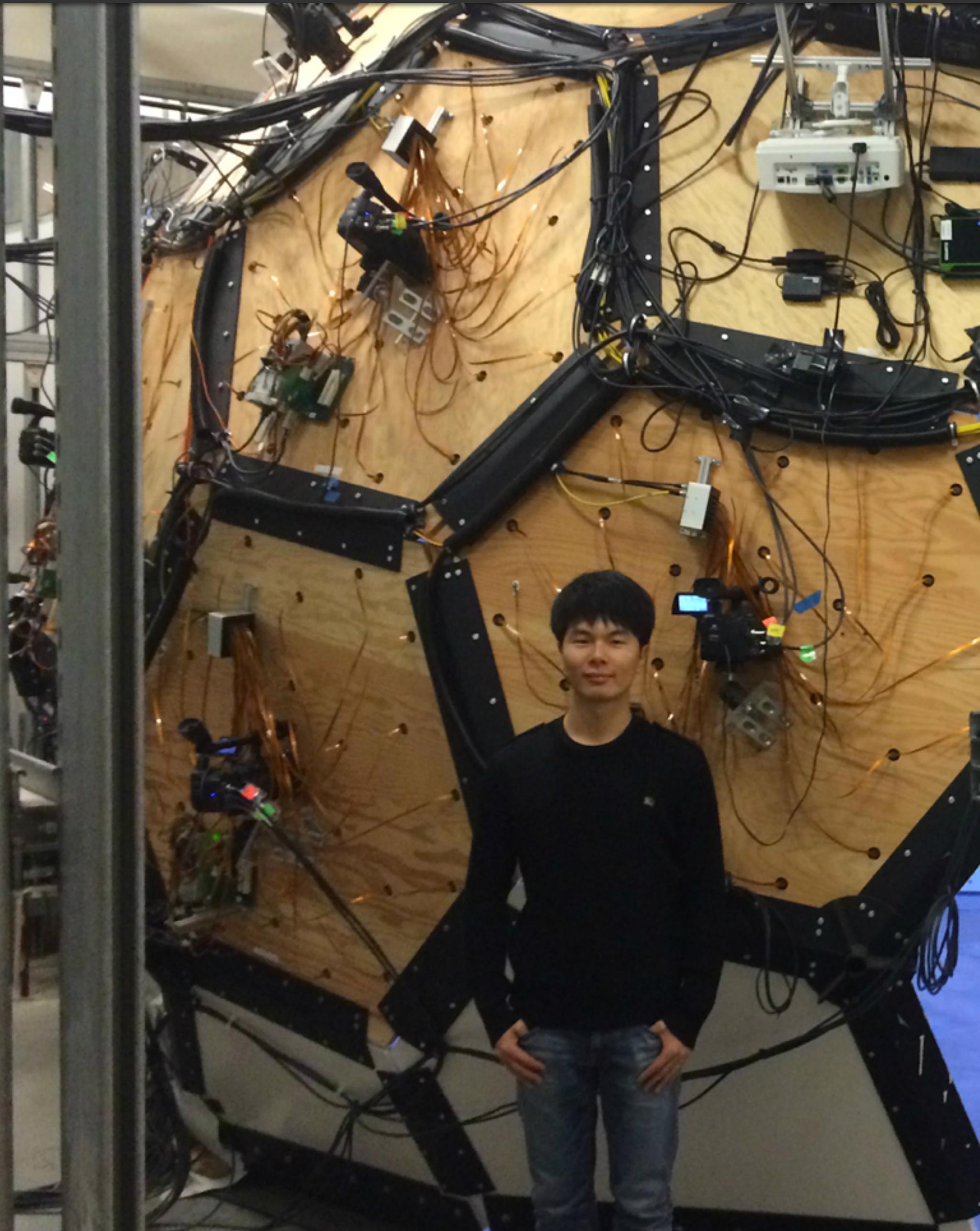
[Courtesy Yaser Sheikh]

Capturing subtle facial expressions



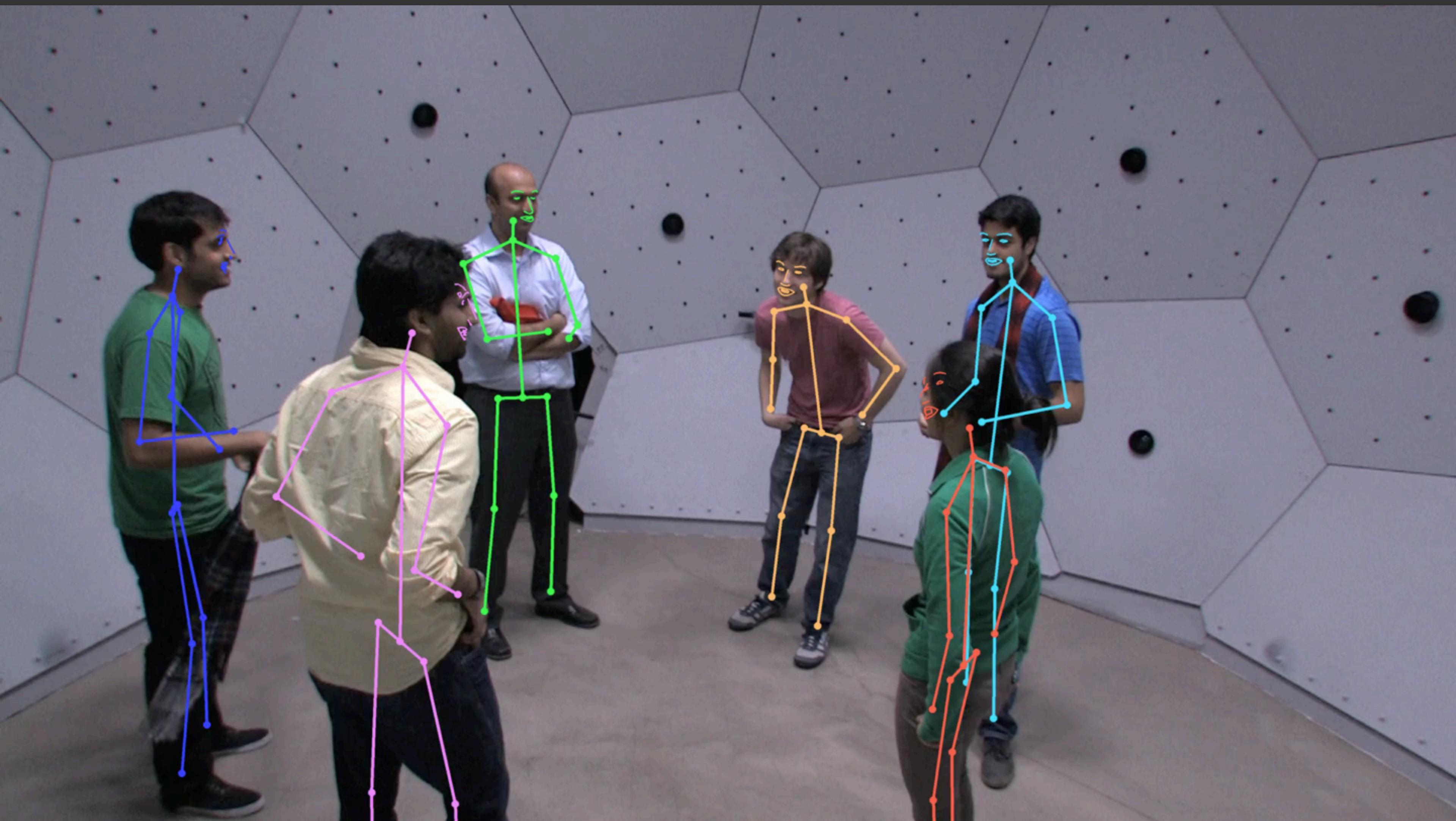
Sensing human social interactions

[Joo 2015]



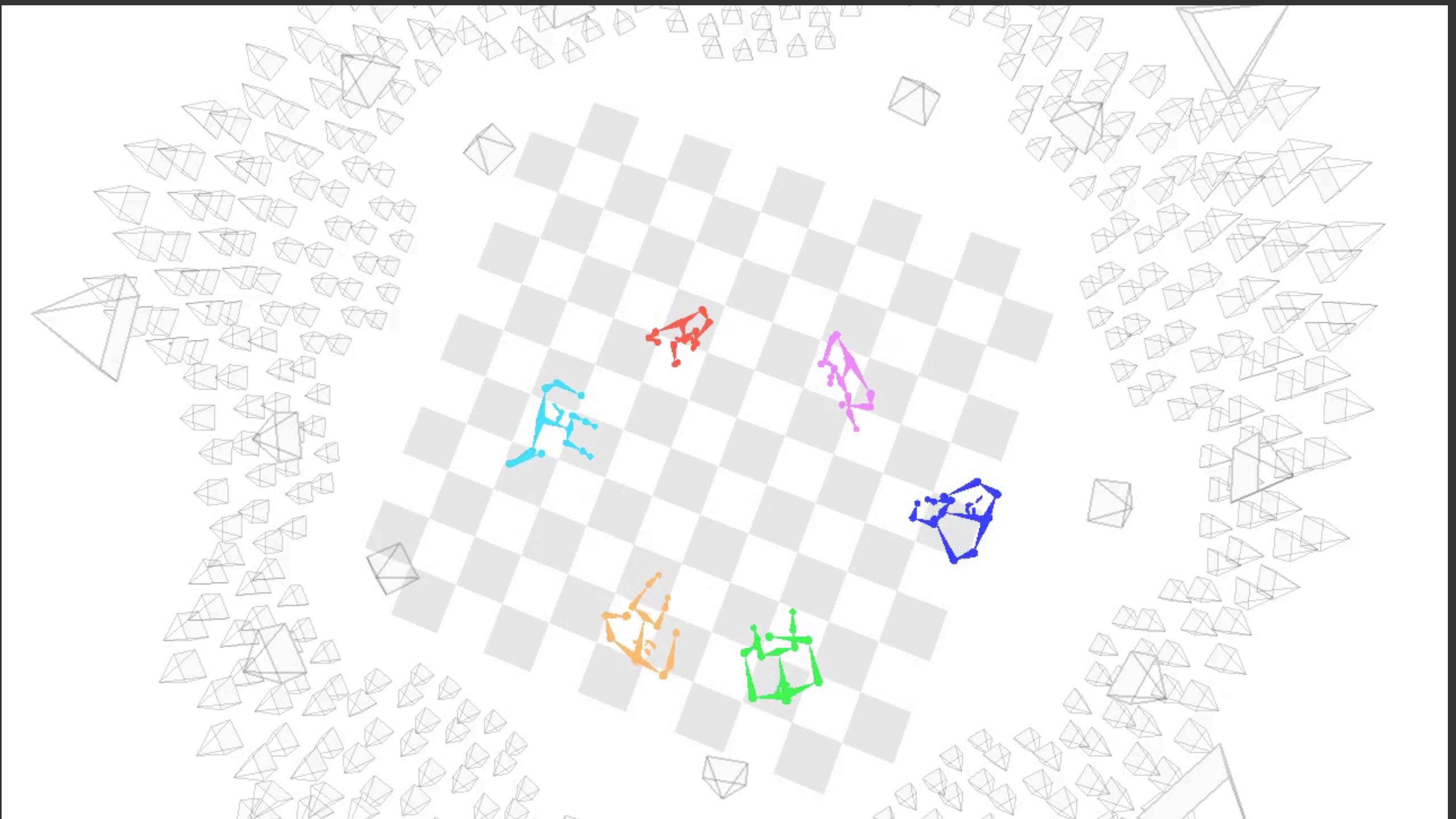
CMU Panoptic Studio
480 video cameras (640 x 480 @ 25fps)
116 GPixel video sensor
(2.9 TPixel /sec)

Capturing social interactions



[Courtesy Yaser Sheikh, Tomas Simon, Hanbyul Joo]

Capturing social interactions



[Courtesy Yaser Sheikh, Tomas Simon, Hanbyul Joo]

What is the latent dimensionality of social signals?

$$\begin{matrix} & \text{BRDF} \\ & (\text{surface appearance}) \\ \downarrow & \\ 5 \text{ million}^1 \times (12 + 100)^2 \times 300 \text{ Hz} \times 3 = 500 \text{ GB/sec} \\ \uparrow & \uparrow \\ \text{3-space} & \text{Sampling rate}^3 \\ \text{coordinates} & \end{matrix}$$

of people

¹ Based on USC-ICT Scan Resolution of Faces

² Kautz et al., "Fast Arbitrary BRDF Shading for Low-Frequency Lighting Using Spherical Harmonics," 2002.

³ Andersson et al., "Sampling frequency and eye-tracking measures: how speed affects durations, latencies, and more", Journal of Eye Movement Research, 2010.

Context is learned over time

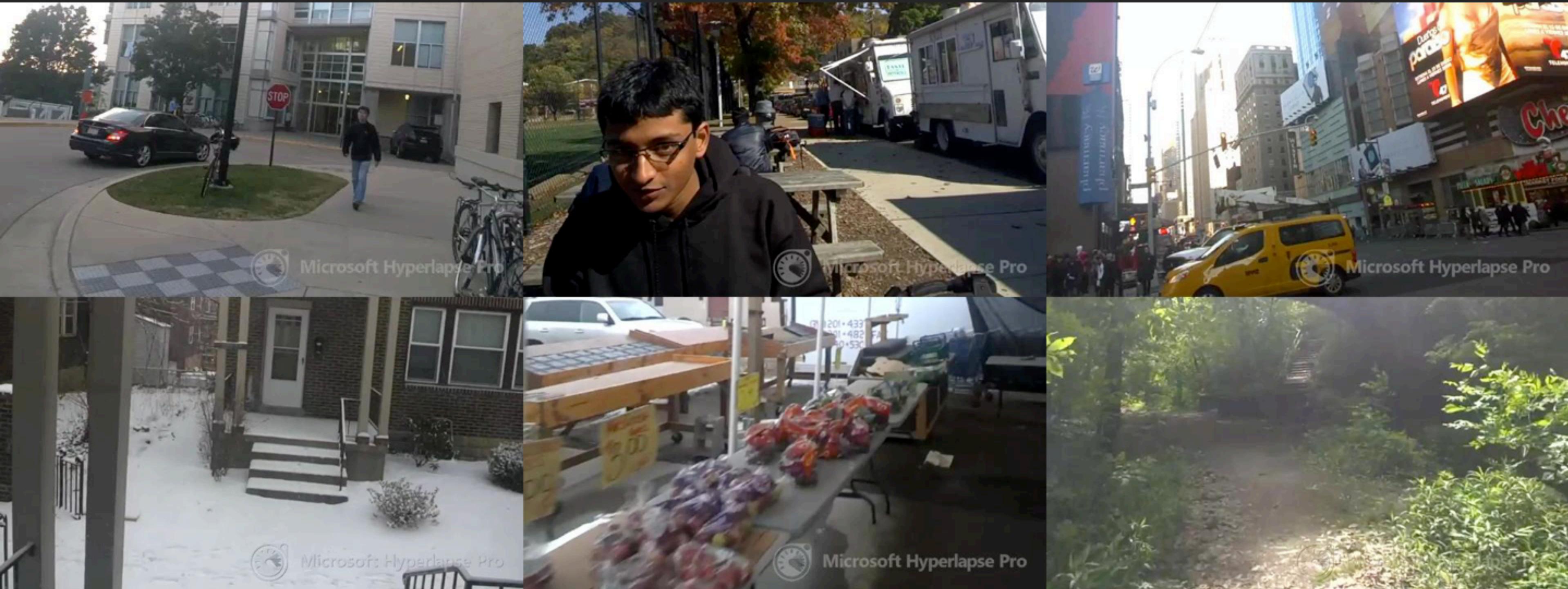
“KrishnaCam” egocentric video dataset



72 hours of recording
over nine months:
(Sep 2014 – May 2015)

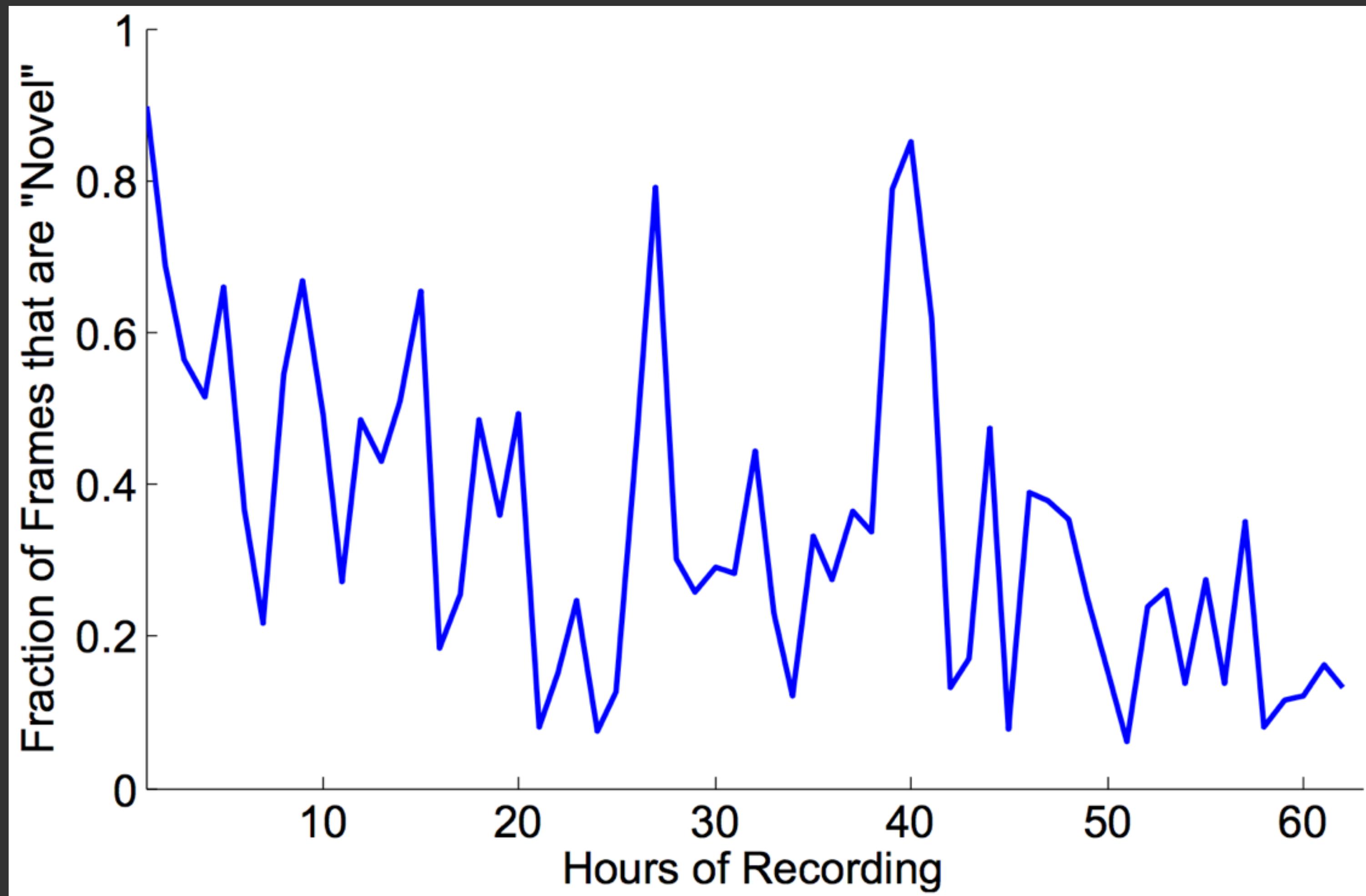
Google Glass





Novel data growth

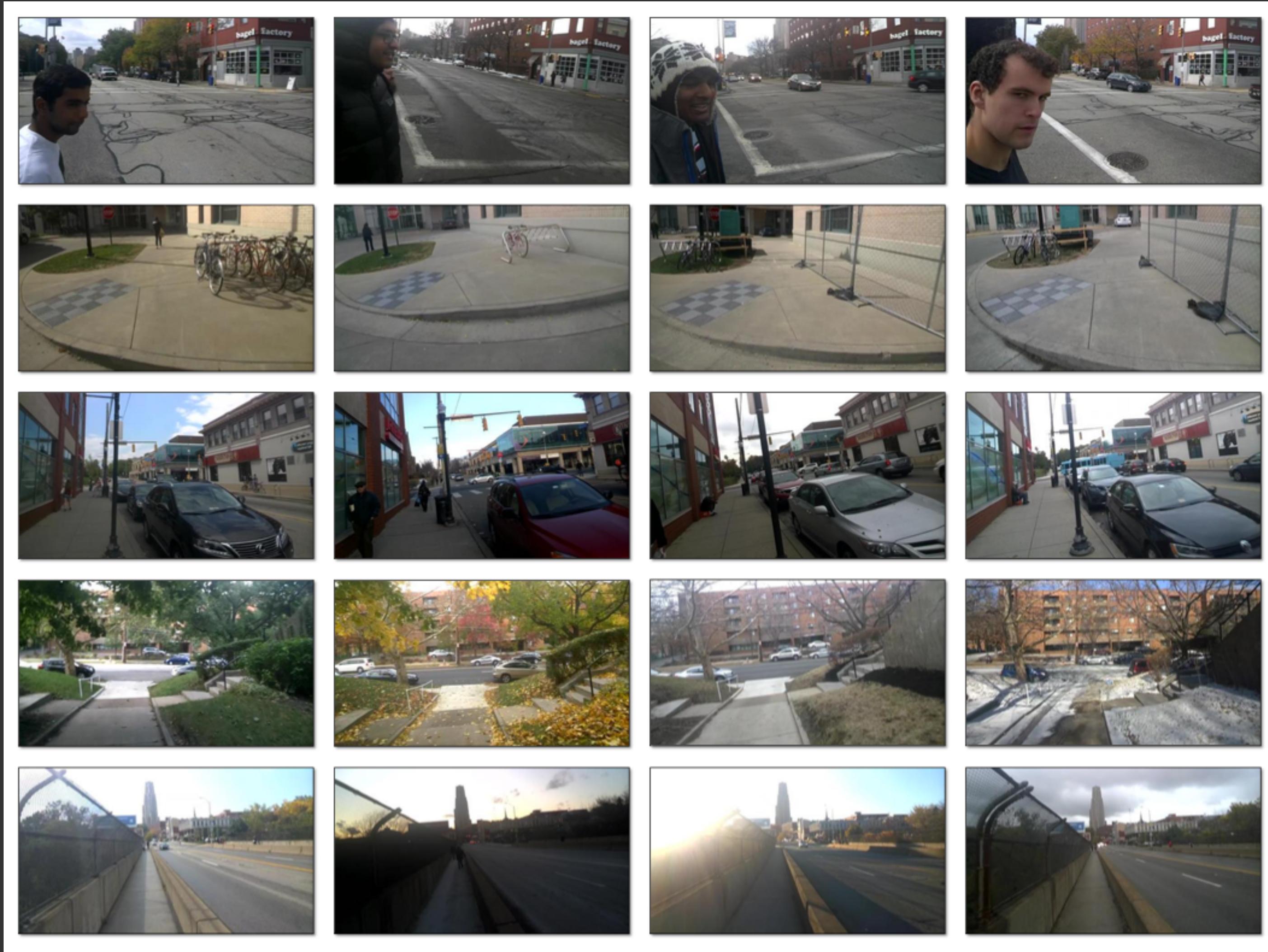
How much new visual data is seen as recording continues?



Similarity = cos distance of MIT Places layer 5 responses (full scene)

"Novel frames" = average distance to top-5 nearest neighbors greater than threshold

How does the world evolve?



1. Change in companion

2. Change in object location (bike rack moved)

3. Change in object (different parked cars)

4. Change in season

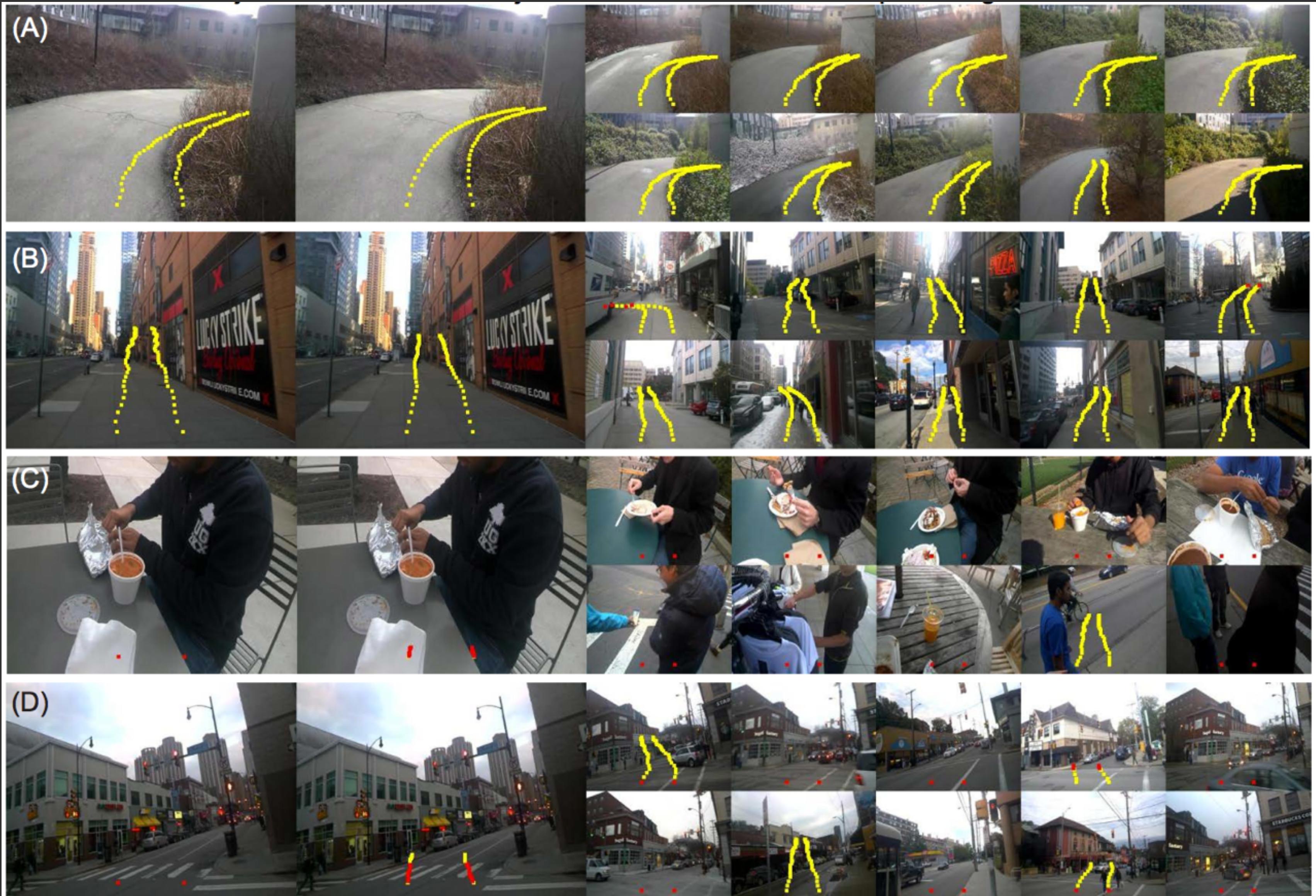
5. Change in time of day (lighting conditions)

Predicting where Krishna will walk next

Current Frame:
Ground Truth Traj

Predicted Traj

Top 10 nearest neighbors to current frame



**Capturing to localize and navigate
(cameras on every vehicle and robot)**

Robot navigation depends on low-latency localization and surrounding object recognition



Under the bonnet

How a self-driving car works

Signals from **GPS (global positioning system)** satellites are combined with readings from tachometers, altimeters and gyroscopes to provide more accurate positioning than is possible with GPS alone

Radar sensor

Ultrasonic sensors may be used to measure the position of objects very close to the vehicle, such as curbs and other vehicles when parking

The information from all of the sensors is analysed by a **central computer** that manipulates the steering, accelerator and brakes. Its software must understand the rules of the road, both formal and informal

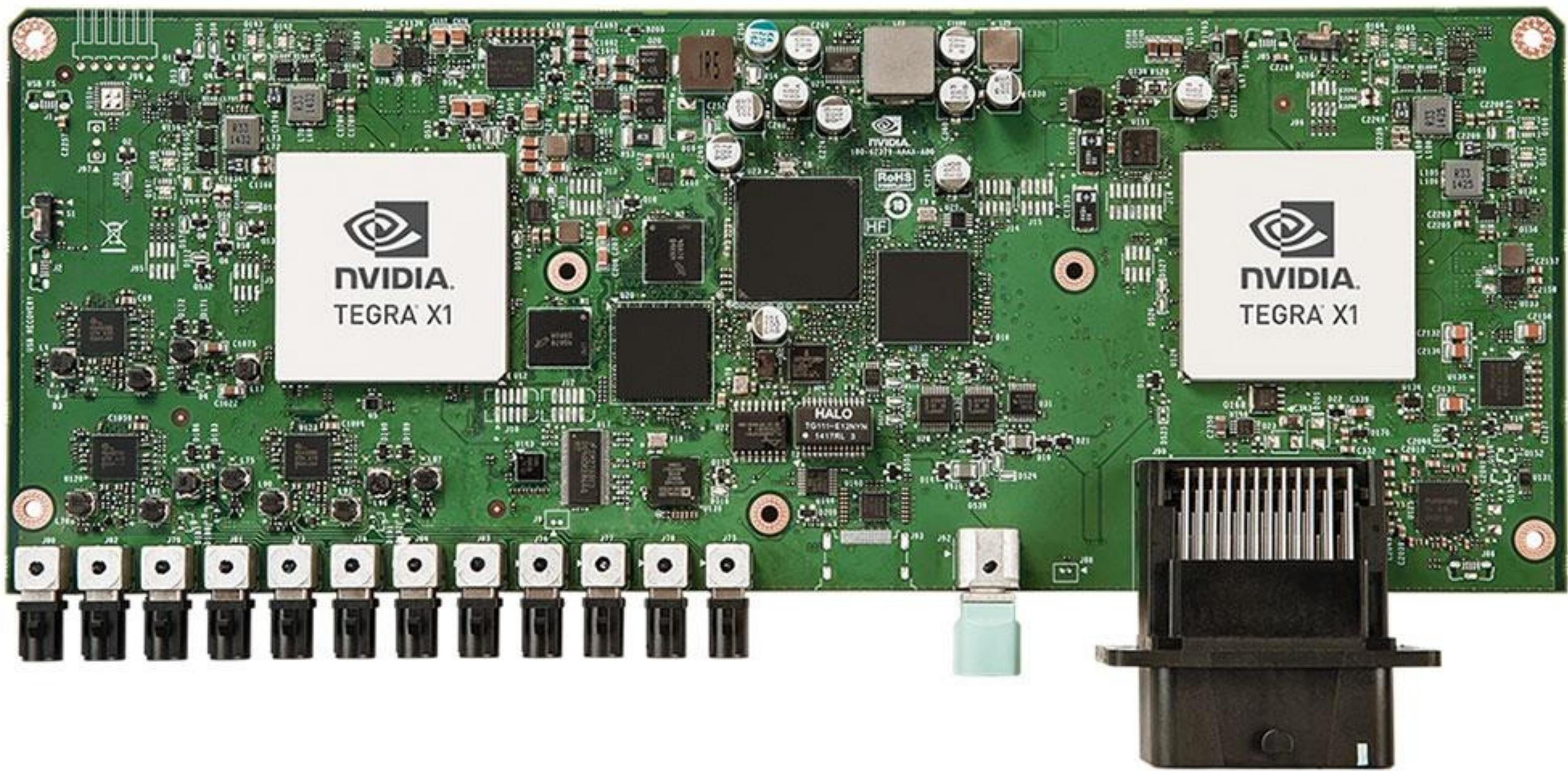
Lidar (light detection and ranging) sensors bounce pulses of light off the surroundings. These are analysed to identify lane markings and the edges of roads

Video cameras detect traffic lights, read road signs, keep track of the position of other vehicles and look out for pedestrians and obstacles on the road

Radar sensors monitor the position of other vehicles nearby. Such sensors are already used in adaptive cruise-control systems

Source: *The Economist*

NVIDIA Drive PX

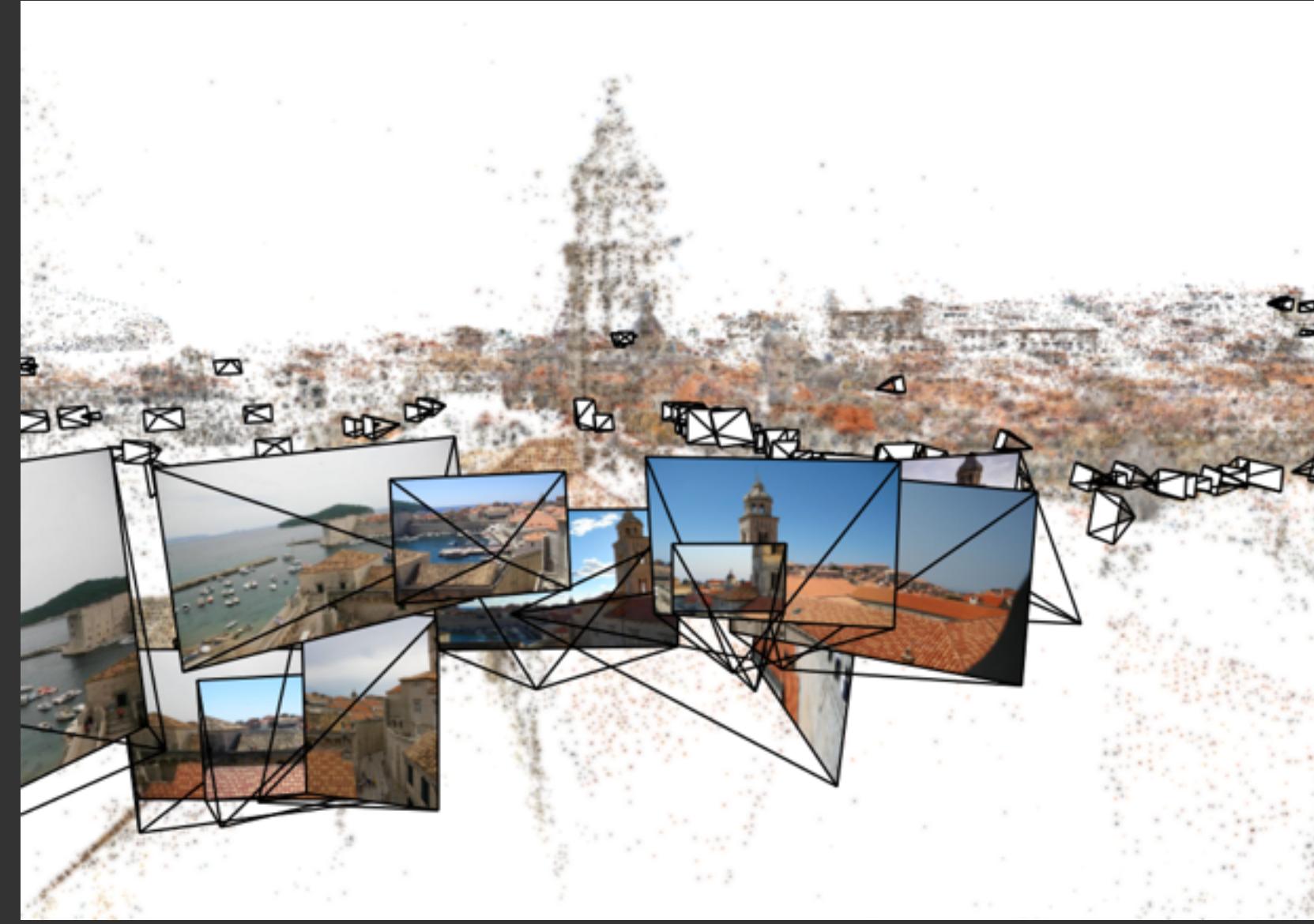


Tegra X1 (1 TFlop fp16 at 1GHz)

AR requires low-latency localization and scene object recognition

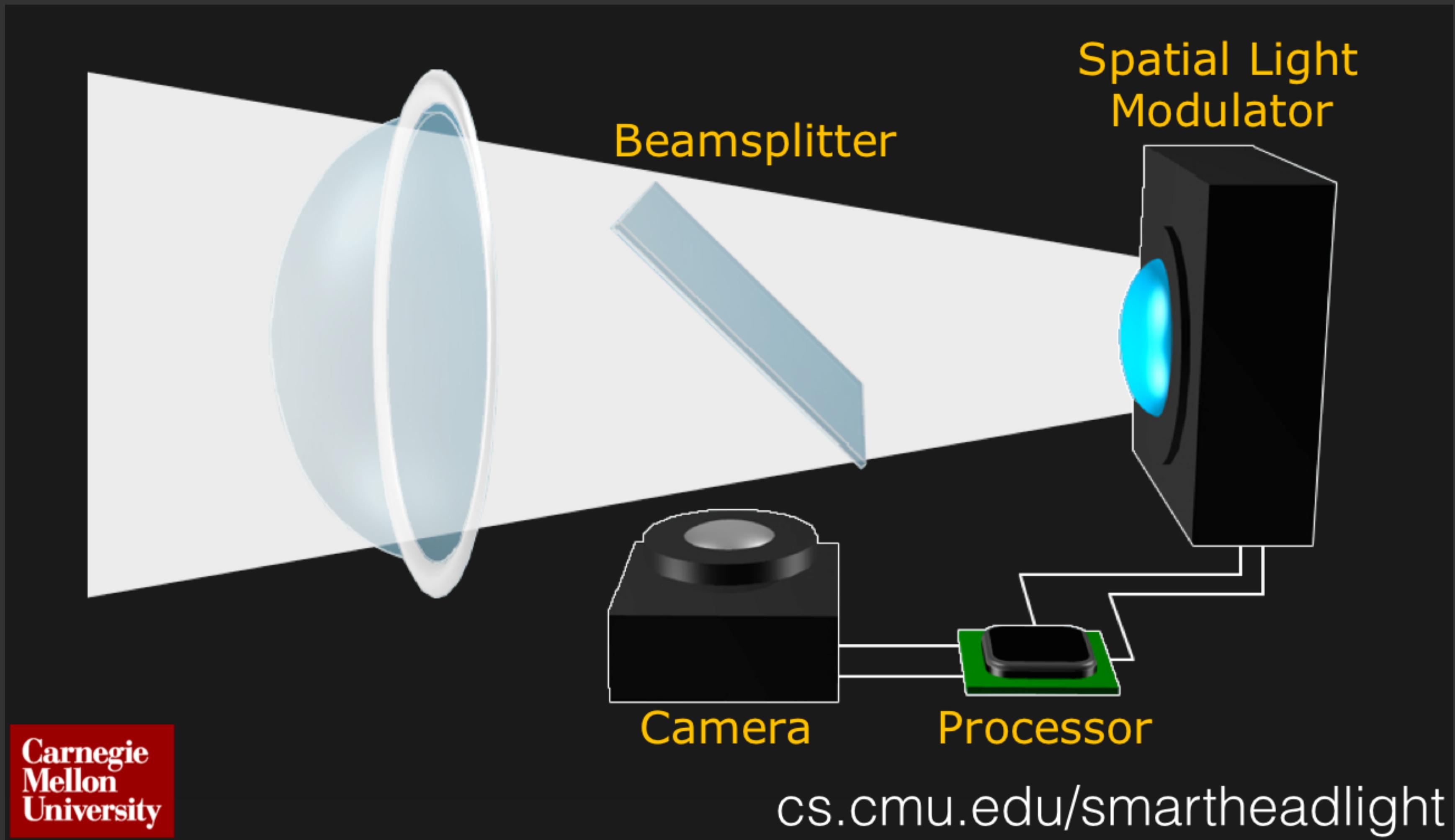


Making “maps”: pervasive 3D construction

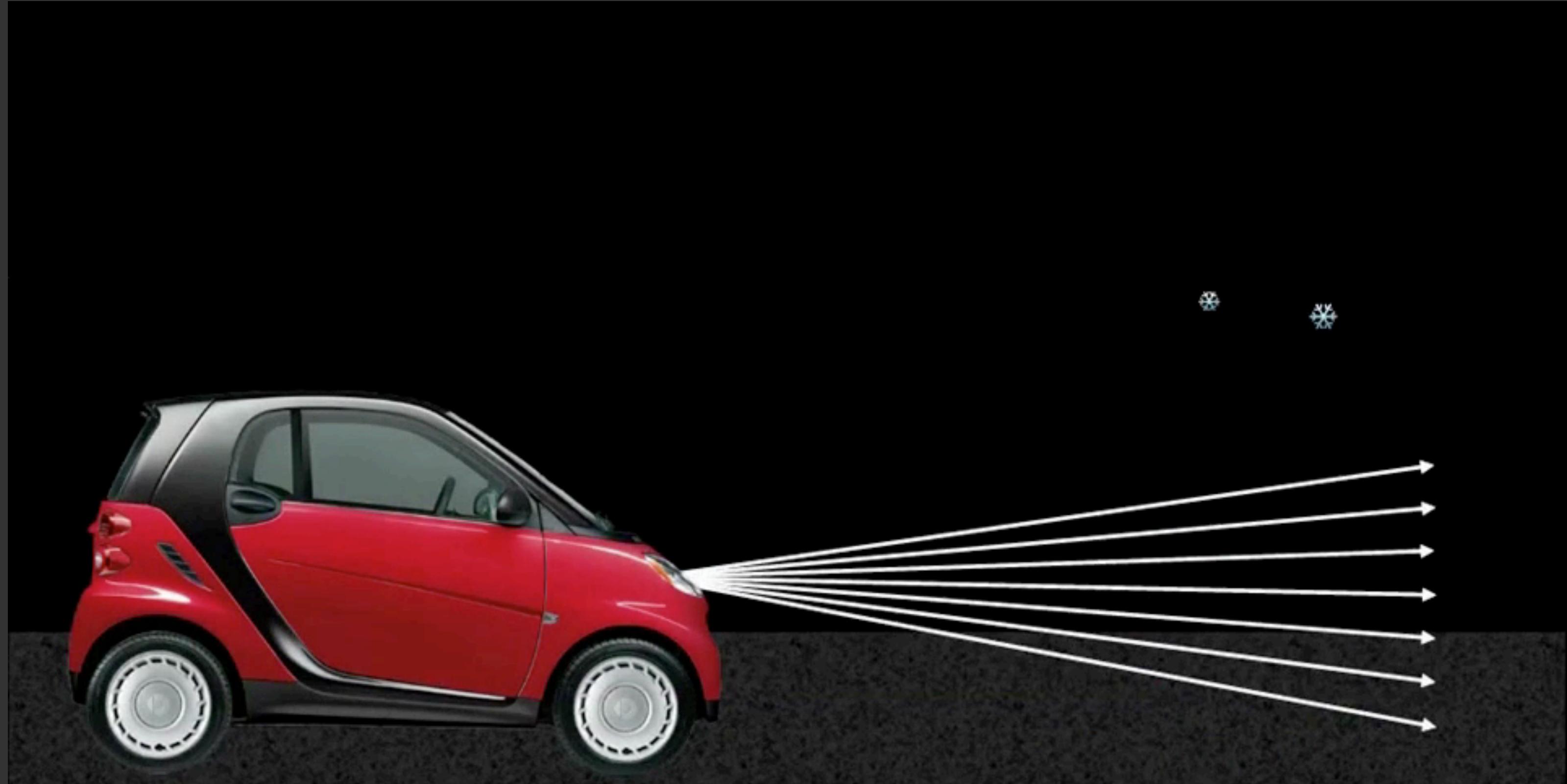




Smart headlight system



Seeing clearly through precipitation



Idea: Stream Light Between Snowflakes

Goal: High Light Throughput and Accuracy

Capturing to understand cities

(Cameras on every street)

(The megacity as the distributed compute/sensing platform of the future)



“Managing urban areas has become one of the most important development challenges of the 21st century. Our success or failure in building sustainable cities will be a major factor in the success of the post-2015 UN development agenda.” - UN Dept. of Economic and Social Affairs

Urban video command center

(Centro de Operações Prefeitura do Rio de Janeiro)



CENTRO DE OPERAÇÕES
PREFEITURA DO RIO

IPLANRIO

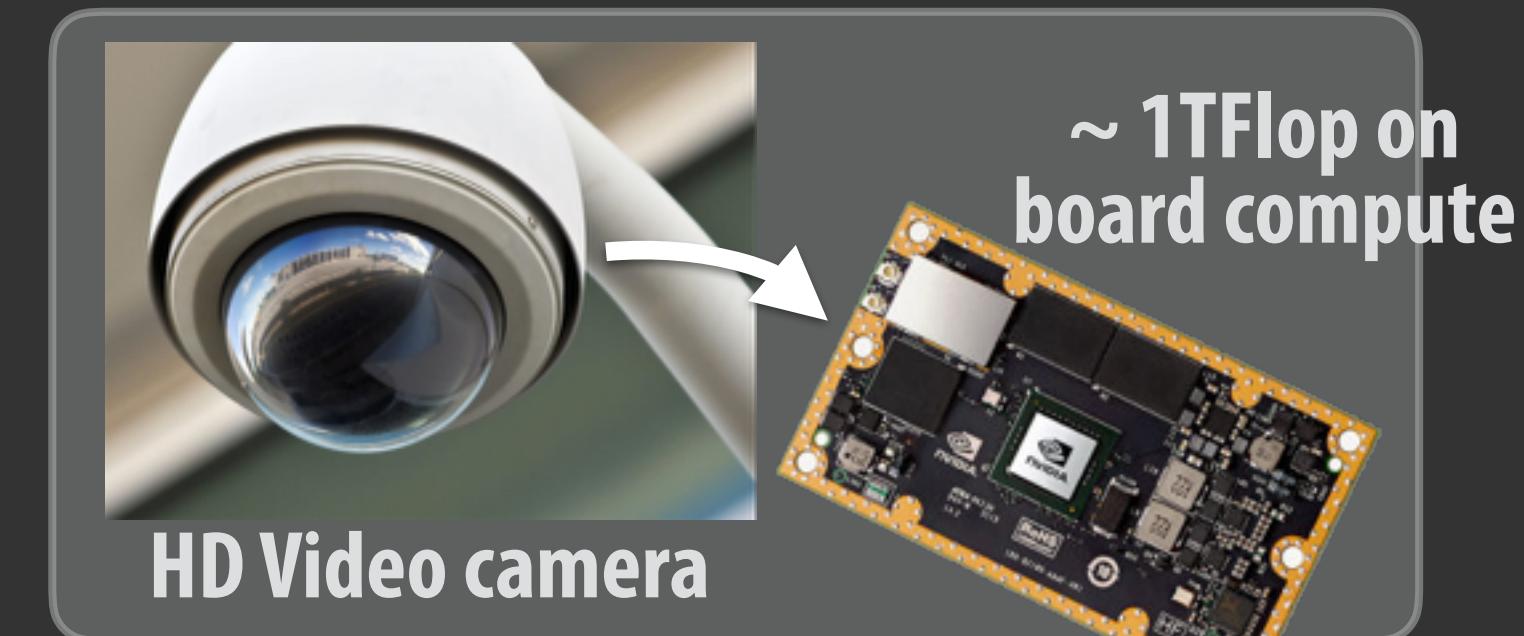
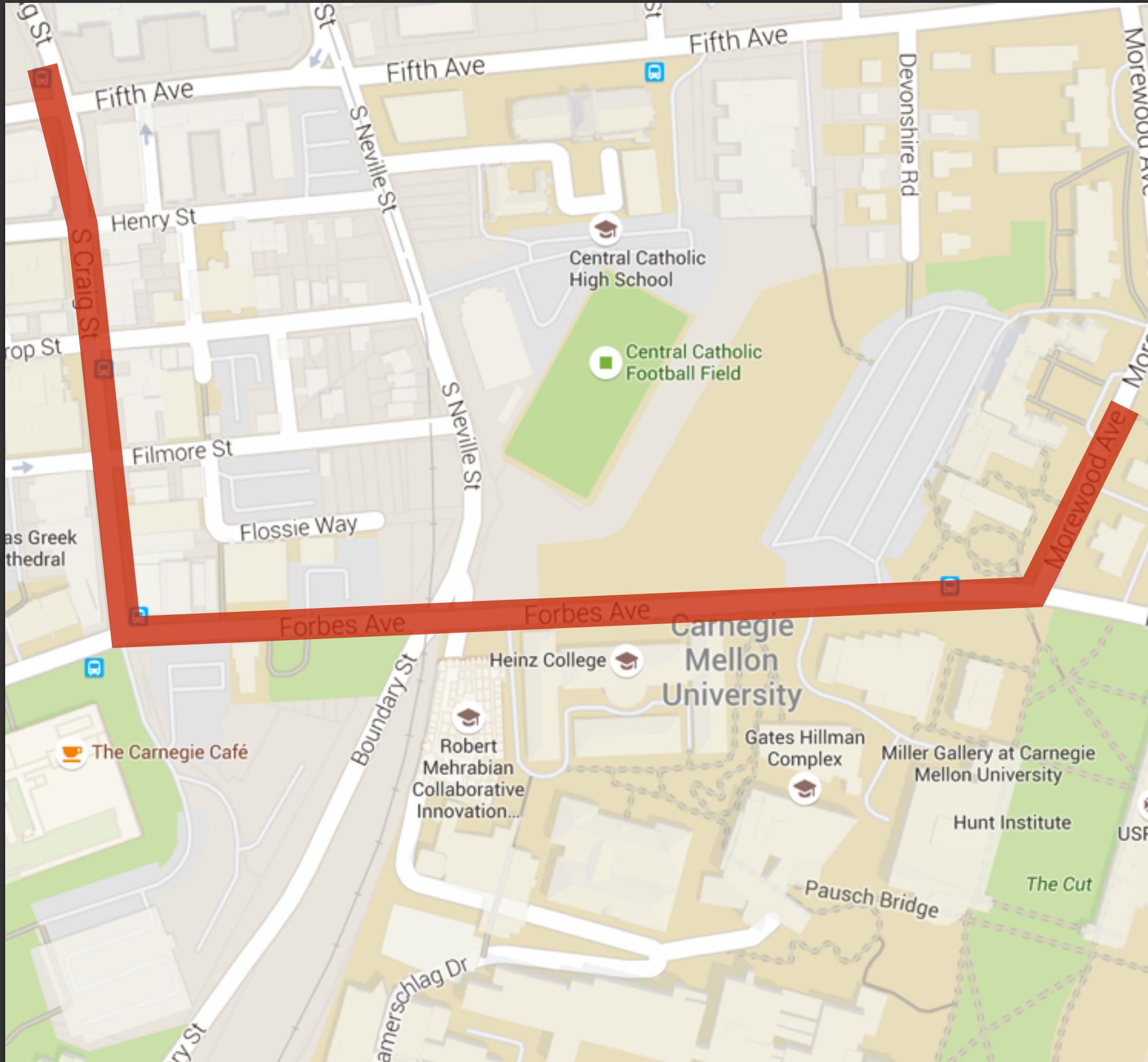
maueill



Urban camera deployments today

- 245M security cameras deployed worldwide (this number includes government owned and private)
 - 6,000 networked cameras in NYC
 - ~500,000 in Beijing [100% public area coverage]
 - 6M in UK, 20M in China
- Purpose is largely to observe and achieve for human query
 - Some ability to perform face / license plate detection, motion detection

Distributed software platform for Pittsburgh-scale video-based data mining and analytics



High speed link



On-campus Parallel
Data Lab Datacenter

1. Use sensing infrastructure to actuate. How can video-based analytics improve city efficiency?
2. How do we build an platform that supports analytics application development for “all cameras in a city”?

Goal: establish viability of city instrumentation to deliver applications that improve efficiency and quality-of-life



~5 sec resolution query-able map of all cars, pedestrians, bicycles, etc.

Open parking spot detection and routing (eliminate circling for parking in greater Pittsburgh)

Postmortem analytics for city planning (How many times was a bike near a bus? Did pedestrians hold up traffic?)

Tracking/localization for autonomous vehicles

Accident or (near accident) detection

Hit-and-run detection (work with insurance companies)

Infrastructure monitoring: pot-hole detection, frozen street detection (salt truck allocation)

Air-quality monitoring

Watch my kids walk home alone after school...

Testbed for addressing interrelated technical, political, and privacy issues



Edge-to-datacenter distribution of computation (scheduling applications across the datacenter and to the edge)

Multi-tenancy near the image sensor (multiple applications must share sensor feeds)

First-class DBMS support for visual computing data

Programming systems for expressing video analysis applications for this infrastructure (“how to program a city”)

New computer vision models for attention and compression (leveraging history and priors to reduce datacenter ingest)

New representations for images and videos that preserve privacy (what information is acceptable to leave the camera? Blurred faces? Features?)

Working with local city government to establish policy and protocol as a research output.

The world in 2030

The world in 2030

- **8.5 billion people** [UN estimate]
- **61% urban (41 “megacities” of 10M people or more)** [UN estimate]
- **2 billion cars** [Sperling 2009]
- **700 - 1.12B streaming security video cameras**
 - Extrapolation from 245M in 2014, for growth between 7-10% [IHS]
- **Assume 8K (7680 x 4320) stereo sources (2 x 33 MPixel image)**
- **Total continuous capture capability of the world:**
 - **25.6B video streams**
 - **1.7×10^{18} pixels \approx 2 quintillion pixels (2 exapixels)**

The world in 2030

- Total continuous capture capability of the world:
 - 25.6B streams (assumed 8K stereo)
- Consider evaluating a modern object-detection deep neural network (GoogLeNet) on every frame from these streams at 30 fps $\approx 10^{12}$ images/sec
 - Today: Tegra X1 fp16: 12 images/sec/watt on tiny 224x224 images [NVIDIA]
 - Let's (naively) multiply per frame cost by 100 to account for larger image size
 - With today's technology: 10^{13} Watts
 - Estimated world's power consumption in 2013: 10^{13} Watts

Final thoughts

- Computer graphics has always involved a healthy interaction between architecture, programming systems, and algorithms
 - Domain focus has been exceptionally useful for vertical thought
 - Willing to throw out old and re-engineer software (new hardware enables programs that haven't been written yet!)
 - Architects should know the algorithms well, and influence them!
- Visual computing has always challenged computer systems by its desire to simulate/synthesize complex visual information
- Next 1-2 decades: interpreting the worldwide visual signal
 - Acquiring and modeling everything humans would see, to enable computers to interpret and analyze
 - We will continue to take every op (op/Watt) you can give us

Thank you

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