

Professor Anil Kokaram

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5C1 Motion Picture Engineering

Digital Video Engineering
Engineering for Moving Pictures

Career

- 1986 – 1989 Electrical and Information Sciences @ Cambridge Uni
- 1989 – 1993 Ph.D Cambridge University DSP Group www-sigproc.eng.cam.ac.uk
- 1993 – 1997 Research Fellow Cambridge Uni and Churchill College Cambridge
- 1998 – 2011 Professor, www.sigmedia.tv, EE TCD
- ...

Research in

Digital Signal Processing

Digital Video Processing

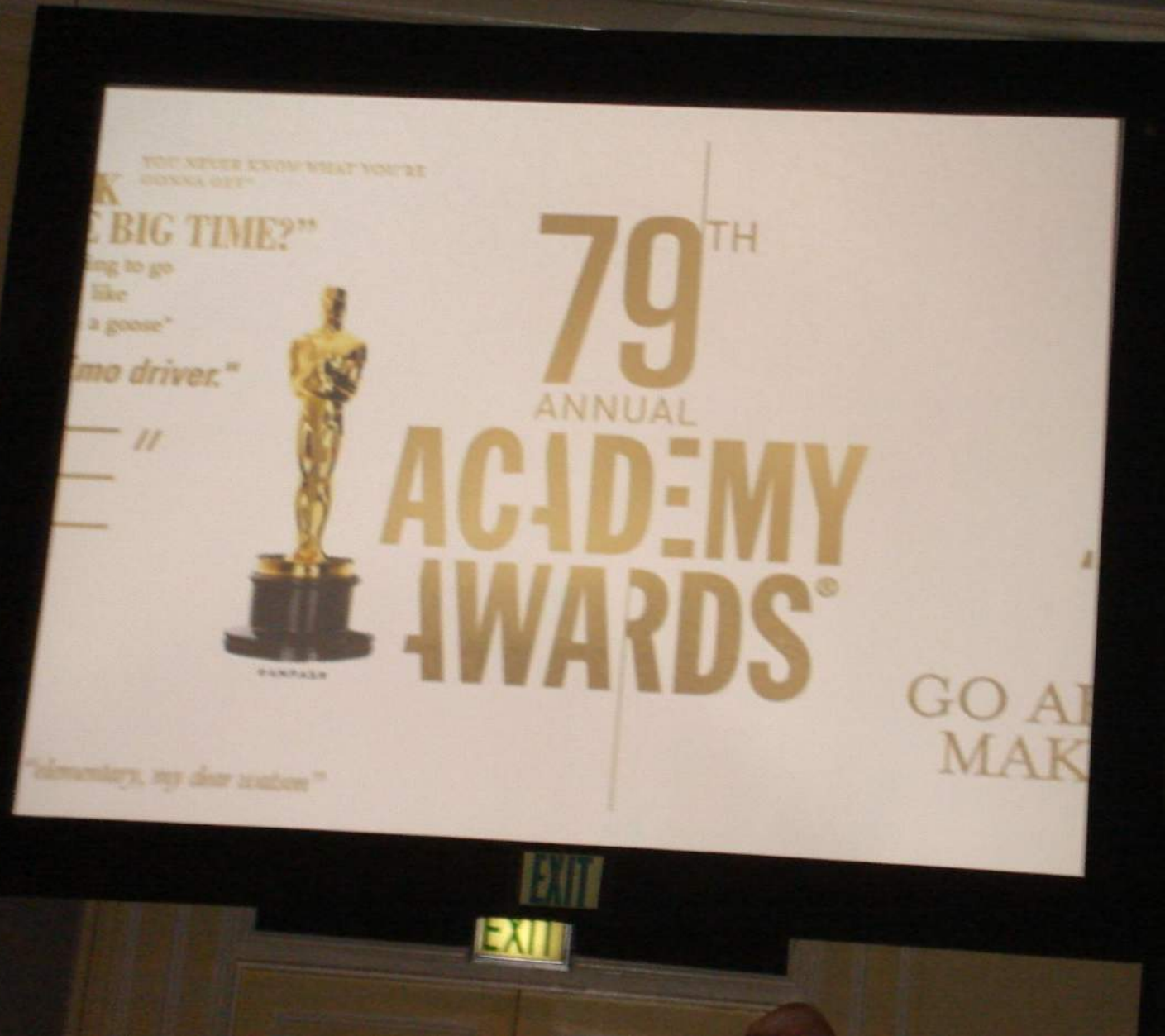
Motion Picture Restoration

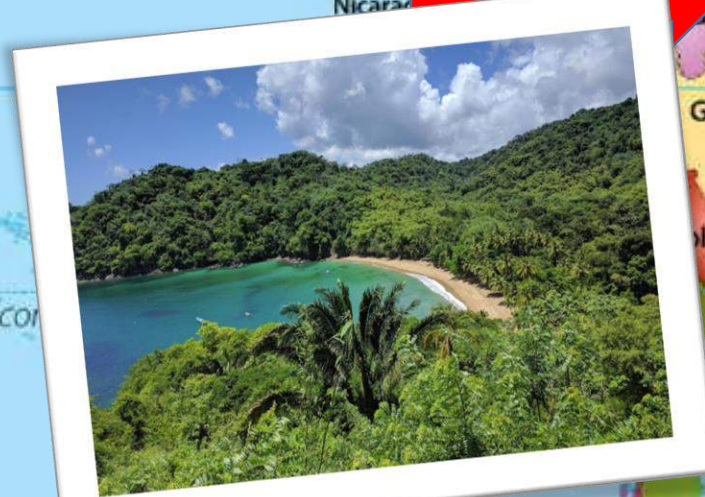
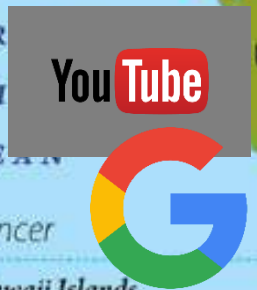
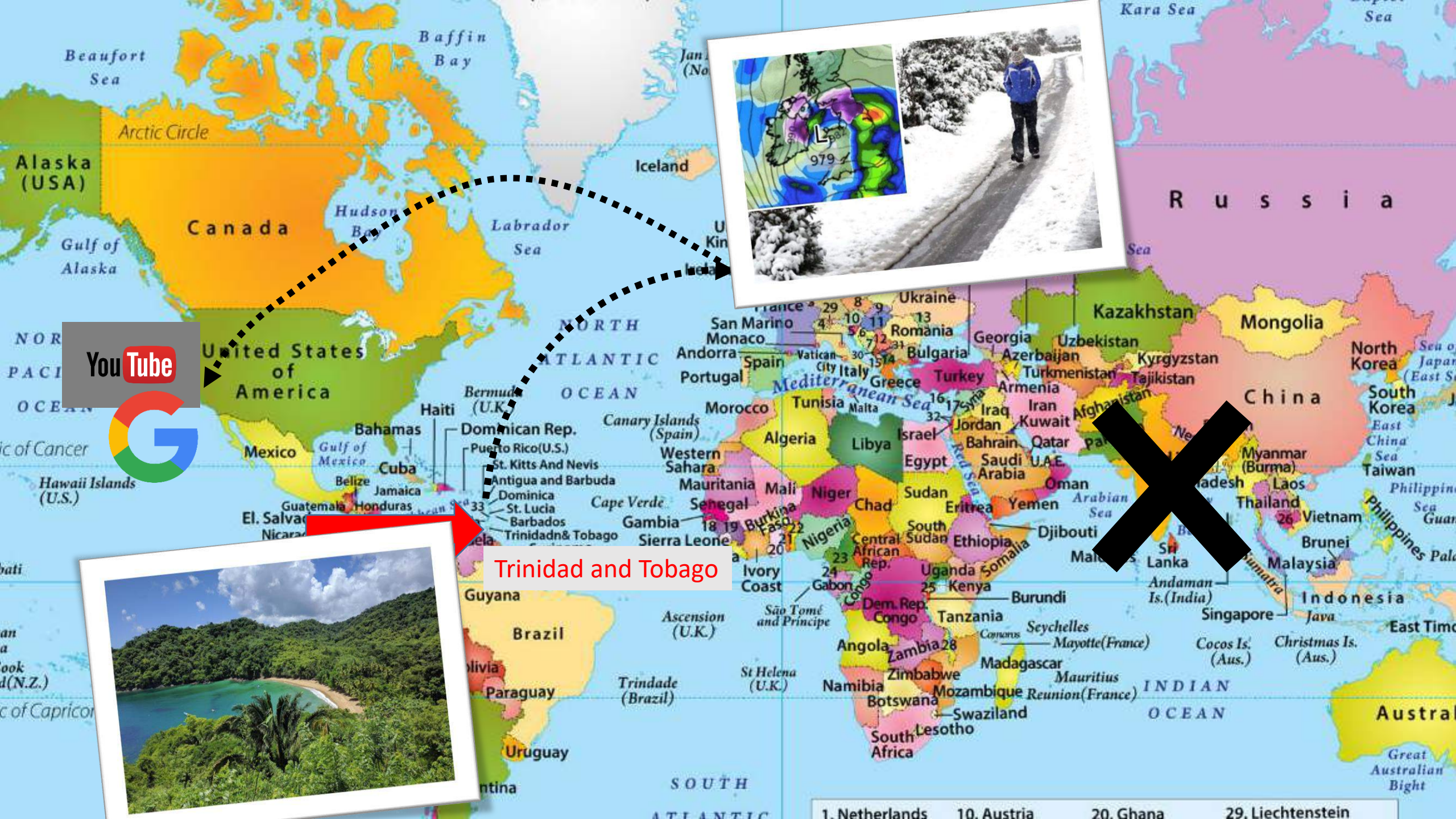
Anything to do with pictures and sound or interesting data

www.sigmedia.tv

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- ➔ • 2007 Academy Award for Motion Estimation FX





Trinidad and Tobago

- | | | | |
|----------------|-------------|-----------|-------------------|
| 1. Netherlands | 10. Austria | 20. Ghana | 29. Liechtenstein |
|----------------|-------------|-----------|-------------------|

GREEN PARROT PICTURES

Restoration Enhancement Conditioning



BBC News - Google buys Parrot Pictures to improve YouTube quality - Windows Internet Explorer

http://www.bbc.co.uk/news/technology-12757153

File Edit View Favorites Tools Help

BBC News - Google buys Parrot Pictures to impro...

BBC News

Home UK Africa Asia-Pac Europe Latin America Mid-East South Asia US & Canada Business Health Sci/Environment Tech Entertainment Video

16 March 2011 Last updated at 08:29 GMT

Google buys Parrot Pictures to improve YouTube quality

Google has bought Irish company Green Parrot Pictures in a bid to improve the quality of video uploaded to YouTube.

The Dublin-based firm specialises in image processing to improve, for example, sharpness and camera shake.

Its technology has already been used by

GREEN PARROT PICTURES
Restoration Enhancement Conditioning

Leaders in Video Motion

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and
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Google's YouTube buys digital video technology firm Green Parrot Pictures | The Australian - Windows Internet Explorer

http://www.theaustralian.com.au/business/news/google-youtube-buys-digital-video-technology-firm-green-parrot-pictures/story-rtty80e-122

Google

Google's YouTube buys digital video technology firm Green Parrot Pictures

Business and enjoying ch Benecol

THE AUSTRALIAN NATIONAL AFFAIRS BUSINESS AUSTRALIAN IT HIGHER EDUCATION VIDEO CARS JOBS CLASSIFIEDS NEWS NETWORK

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Dow Jones +1.06 12381.1104 at 15:25 Oil -0.26 109.66 at 15:25 ASX 200 -72.5 4098.7 at 16:40 AUD/USD -0.0 1.0469 at 15:25

Google's YouTube buys digital video technology firm Green Parrot Pictures

Scott Morrison | Dow Jones Newswires | March 16, 2011 8:40AM

Recommend One recommendation. Sign up to see what your friends recommend. 6 Tweets 2 Shares

GOOGLE'S YouTube said today it has acquired digital video technology maker Green Parrot Pictures for an undisclosed amount.

The Irish start-up's technology will enable YouTube to improve the quality of jerky, blurry or unsteady videos as they are being uploaded to the site, Jeremy Doig, a director for Google's video technology, said in a blog post.

Mr Doig said that with 35 hours of video uploaded to YouTube every minute from people all

Who do you become in nature?
The Curious You?
The Playful You?
The Creative You?

Find out now! DiscoverTheForest.org

Waiting for https://www.facebook.com/plugins/like.php?action=recommend&api_key=14060532623707&channel_u...

http://blogs.forbes.com/ericavitz/2011/03/15/google-buys-video-technology-company-green-parrot-pictures/

Google Buys Video Technology Company Green Parrot Pictures - The Tech Trade - For - Windows Internet Explorer

http://blogs.forbes.com/ericavitz/2011/03/15/google-buys-video-technology-company-green-parrot-pictures/

Google Buys Video Technology Company Green Parrot Pictures

3647 15 2011 - 6:19 pm | 2,447 views | 0 recommendations | 1 comment

Google (GOOG) has acquired Green Parrot Pictures, a Dublin, Ireland-based provider of digital video editing tools and services.

Green Parrot announced the deal on its Web site.

44 We're excited to announce that Green Parrot Pictures has been acquired by Google!

We've enjoyed collaborating with organizations in the film industry as well as all our clients, and we'd like to thank everyone for their generous support.

MY ACTIVITY FEED

Show all activity



ERIC'S NEW POST Update: MagicJack Responds To FCC Ruling In AT&T Case 22 minutes ago

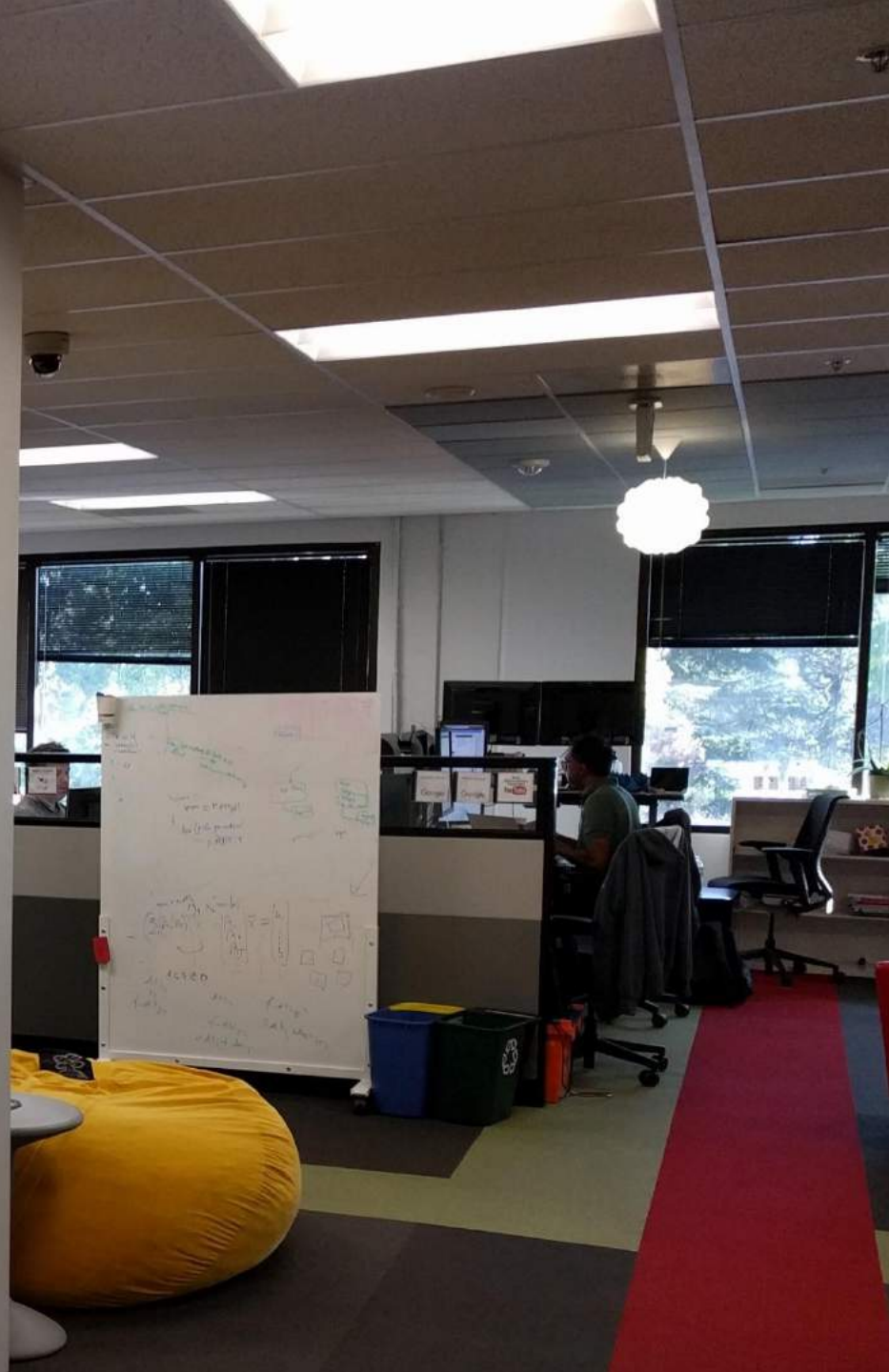
ERIC IS FOLLOWING JESSE DIVNICH The Divnich Dispatch 44 minutes ago

ERIC'S NEW POST Update: MagicJack Responds To FCC Ruling In AT&T Case 1 hour ago

ERIC'S NEW POST Apple: Mixed Views From The 2 hours ago

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-  • 2011 – Trinidad and Tobago Government Honor for “Science and Tech”, Hollywood “Irish Top 50” Honor
-  • 2011 – 2016 Tech Lead @ Google/YouTube “Media Algorithms Group”







ME AT THE 200

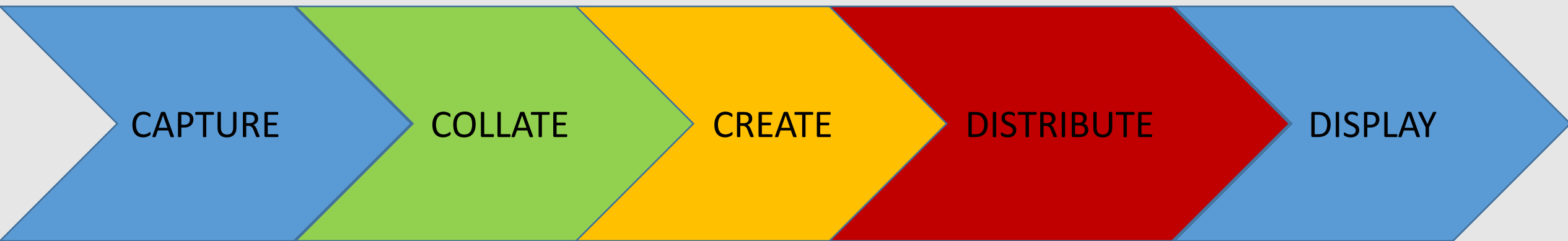
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- ➔ • 2016 – Head EE @ TCD

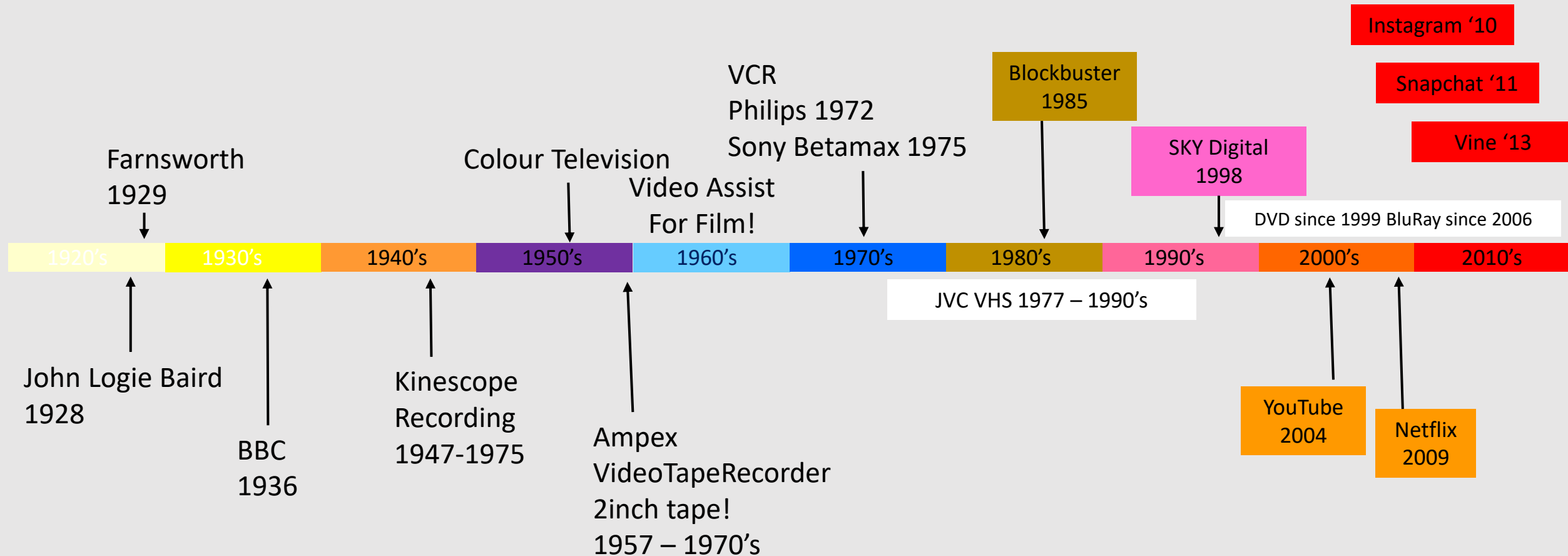
70%
internet is
video

1 Billion
hours on
YouTube

40%
of the world
owns a
smartphone

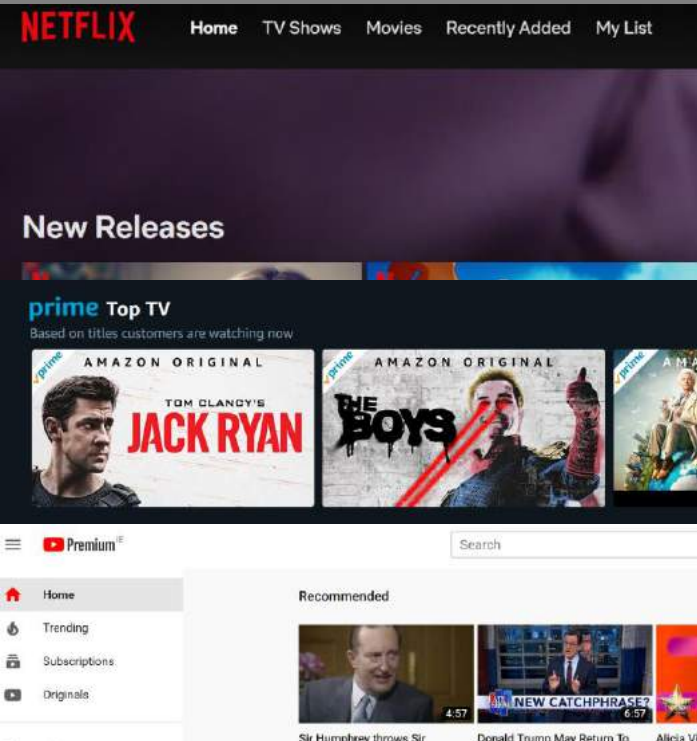


The history of Video/Television

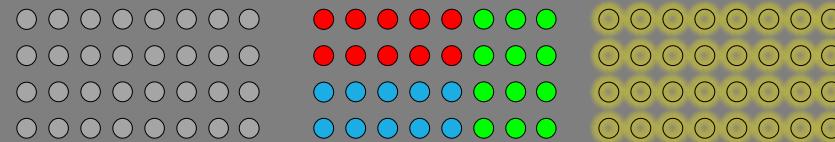


2016 was 80th anniversary Television
1999 was the 100th anniversary of the Cinema

Since 2015



Streaming Media reaches consumers through new picture tech BEFORE traditional broadcasters



**More Pixels (4K, 8K)
Better Pixels (10bits, HDR)
Faster Pixels (60, 120 fps)
Deeper Pixels (RGBX)
Open Royalty Free Video Standards**

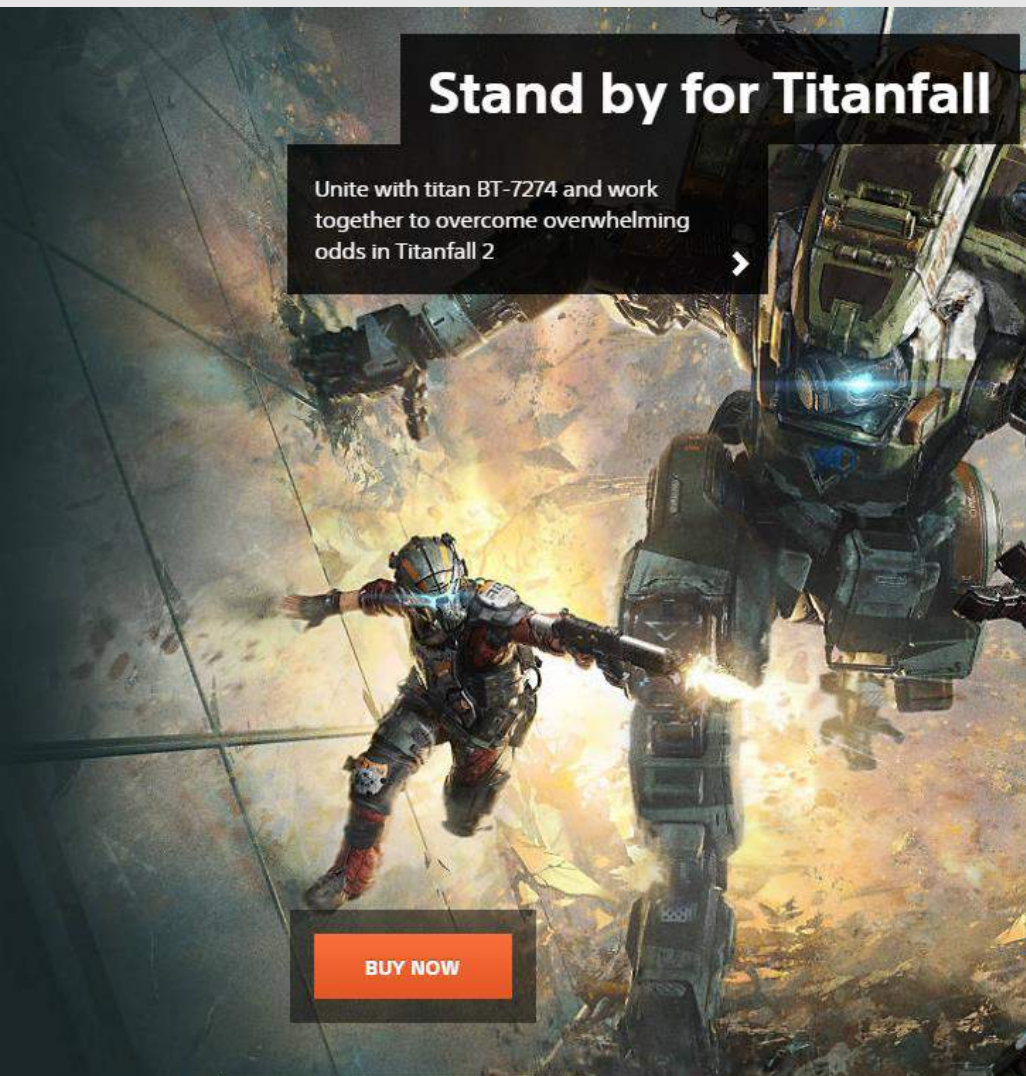


Simple computational tasks are complicated at scale.

We have come full circle.

**Hardware, Compression, "DSP"
Back on the agenda**

Entertainment Drives Technology



10: Gaming Perfected.

NVIDIA's new flagship GeForce® GTX 1080 is the most advanced gaming GPU ever created, driven by the high-performance, power-efficient new NVIDIA Pascal™ architecture.

Pandemic Drives Entertainment

How Zoom defied its critics and became the go-to video conferencing app for surviving the pandemic

By [Jerry Bowles](#) January 18, 2021

7 min reading

SUMMARY: What's next for the insanely popular, death-defying video conferencing phenomenon?

Thanks to thousands of Motion Picture Engineers working on Video Compression Standards, New Technologies for content production and real time video communications The world didn't stop in March 2019. Quite a lot of people were able to continue working and Universities didn't quite close.

in a moment, but first, let's skip to the good parts.

4. Now the race is on for more content

The shutdown in film and television productions, forced by lockdowns this spring, has placed a major focus on how Netflix and its competitors will get hold of new offerings to retain their members.

Netflix crosses 200m subscriber mark in pandemic-boosted year

Covid-19 accelerated 'big shift from linear to streaming', says chief financial officer

© Wed, Jan 20, 2021, 12:58

Updated: Wed, Jan 20, 2021, 15:19



Netflix's Bridgerton, featuring Claudia Jessie (left) and Irish actor Nicola Coughlan, is on track to reach 63 million households, the streamer said. Photograph: Liam Daniel/Netflix

@Netflix in Los Gatos



VIDEO QUALITY EXPERTS GROUP MEETING @SKY CENTRAL, LONDON

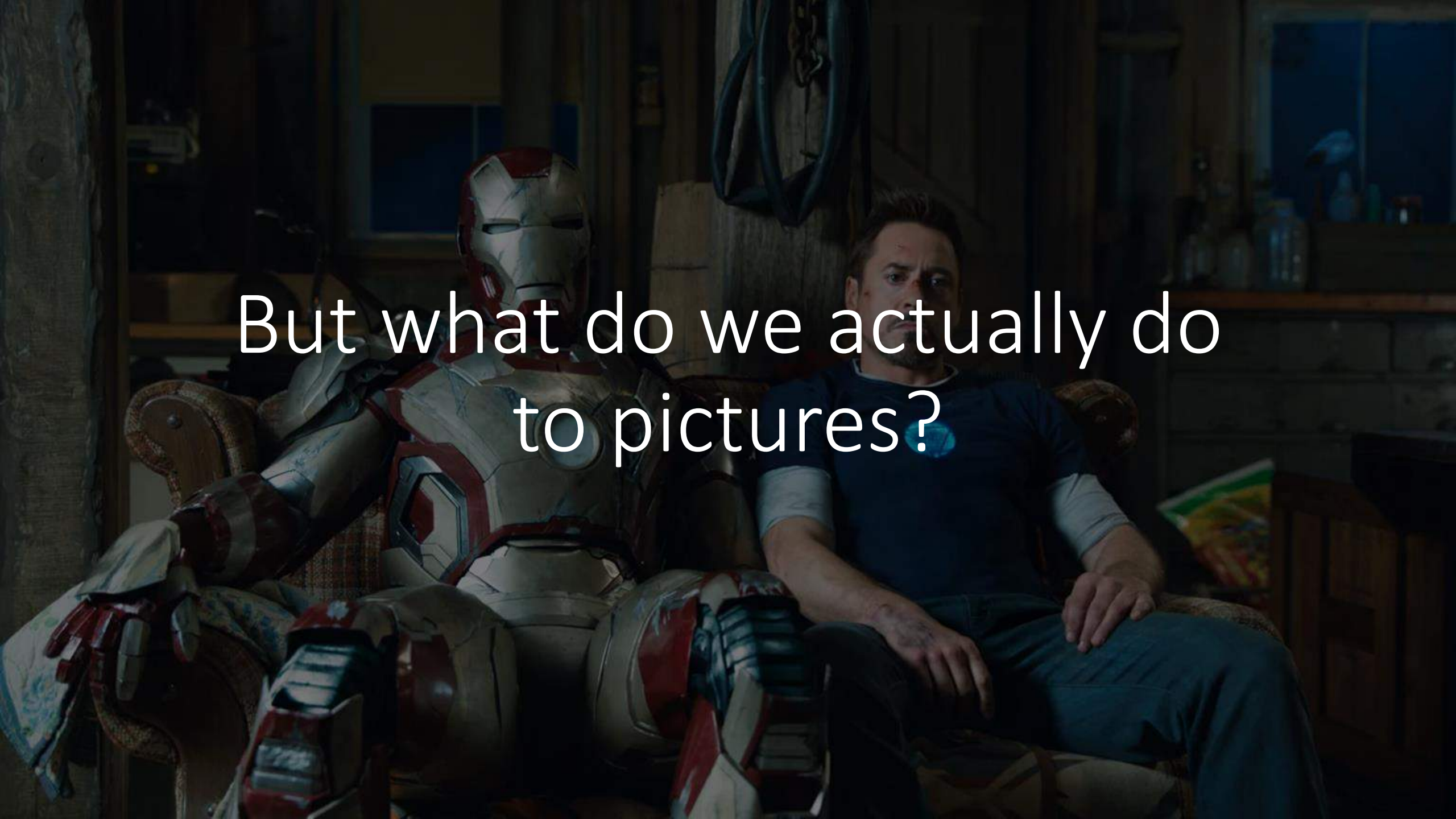


Sunrise
Live from the
Glass Cube
Weekdays 6-10am

sky NEWS

09:57 **sky NEWS** TWO BRITISH TRAVELLERS HAVE BEEN DIA

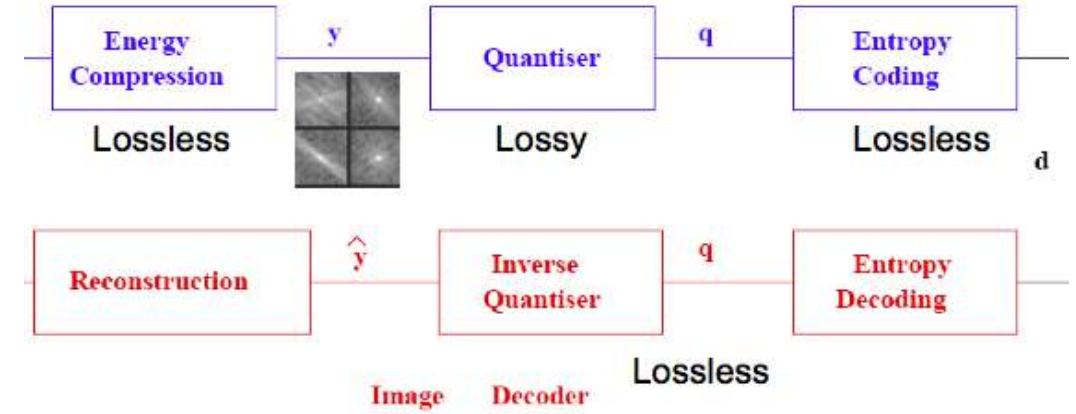
sky
STORE

A scene from the movie Iron Man showing Tony Stark sitting in a wooden chair in his workshop, looking at the Iron Man armor which is also sitting in a chair next to him. The workshop is dimly lit with various tools and equipment visible in the background.

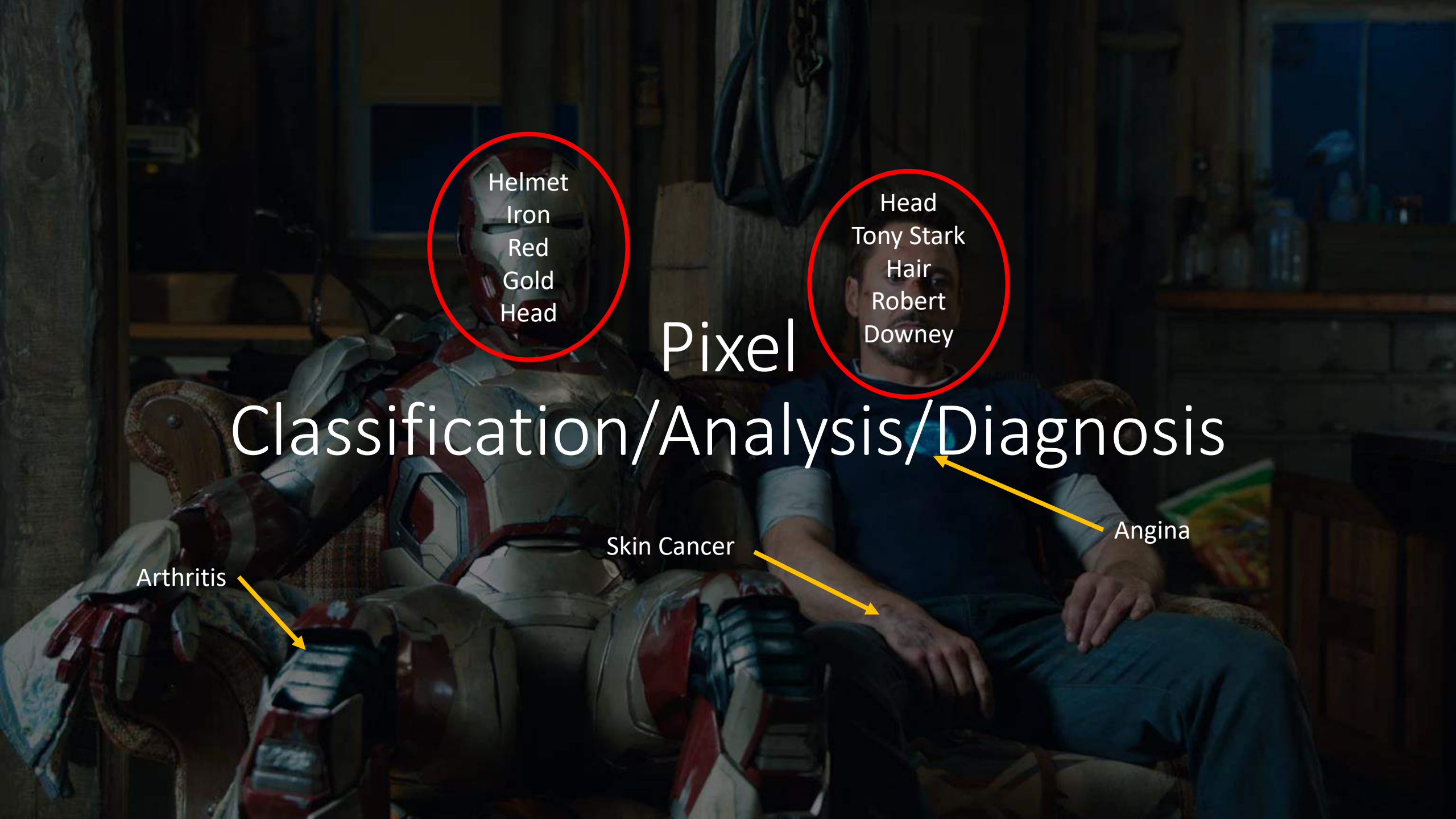
But what do we actually do
to pictures?



Pixel Pushing/Prodding



Pixel Communications



Helmet
Iron
Red
Gold
Head

Head
Tony Stark
Hair
Robert
Downey

Pixel

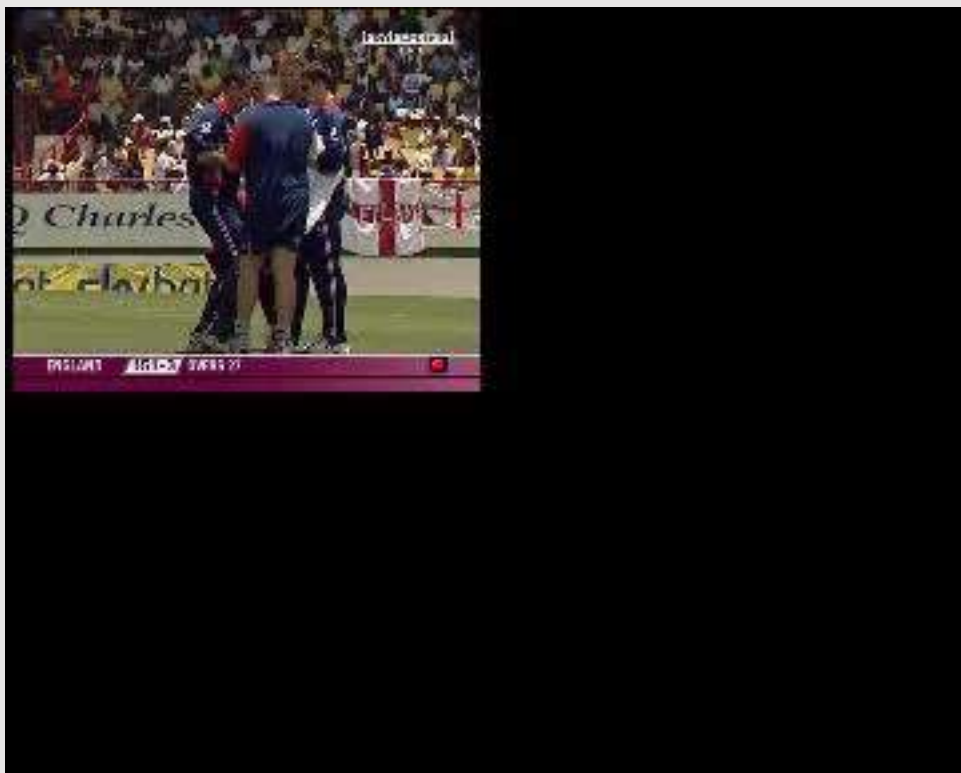
Classification/Analysis/Diagnosis

Arthritis

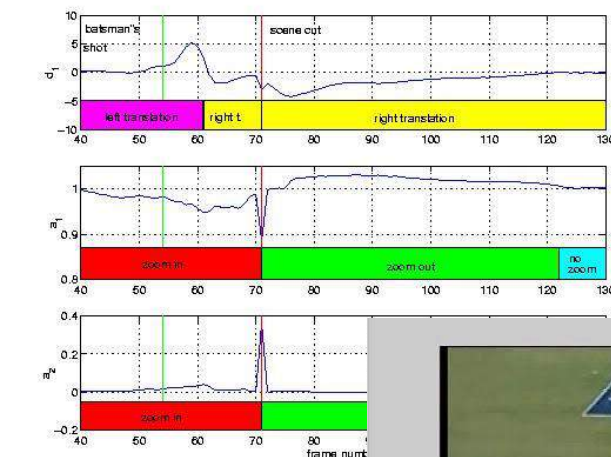
Skin Cancer

Angina

Using audio and video for content analysis



Audio



5C1 : Motion Picture Engineering

- Pixel Pushing (The invisible tools for making better pictures)
- Pixel Communications (Video Streaming)
- Pixel Quality Measurement

This course considers motion pictures in 2D+Time

We don't process the images in the 3D world or do 3D modeling

We also spend a lot of time thinking about actual production and broadcast workflows.

This is not a normal image/video/computer vision course

FOUNDRY.

PRODUCTS

INDUSTRIES

SUPPORT

COMMUNITY

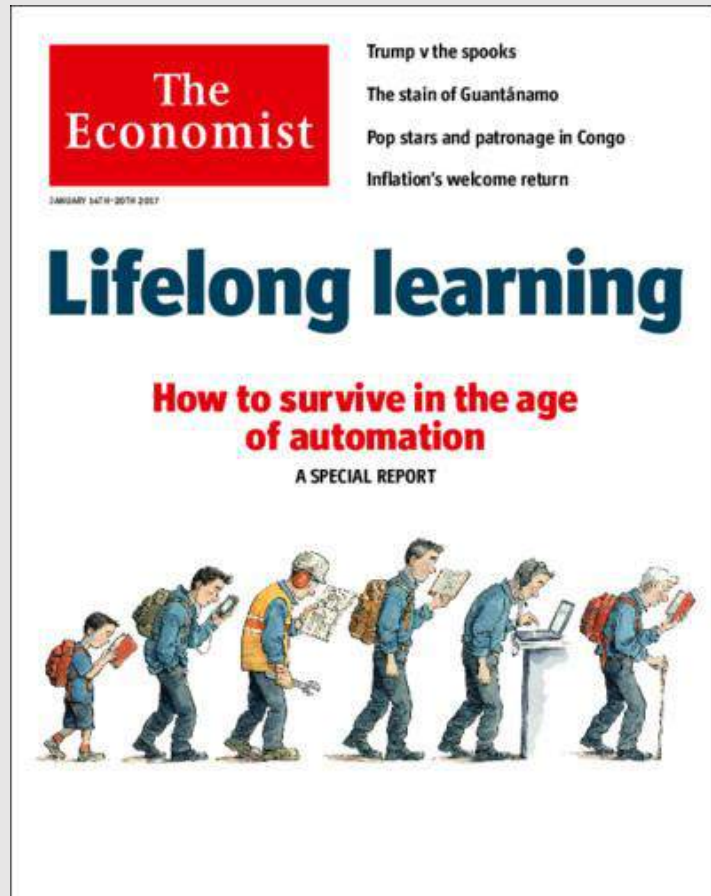


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Motion Picture Engineering Changes....



- Deep Learning is now changing everything !
- You will have to face the realities of more automation and intelligence in everything.
- The next Industrial revolution
- Affects medicine, bioengineering, mechanical engineering, civil engineering, creative technology
- See [this](#)
- In 2019 I was working on a DNN portion of this course .. That got hit in the face on March 2019 and is still on the mat.

An aerial photograph of a dry, cracked landscape. A winding river or stream flows through the center of the image, surrounded by parched, brownish-yellow earth. The cracks in the soil form a complex, branching pattern, resembling a network of veins or a map. The overall tone is somber and desolate, with a dark, overcast sky visible in the upper left corner.

Deep Learning important

but in pixel pushing the data rates and pixel depths are enormous. There is still some way to go...

Course Content

- Week 1 Introduction + Colour Segmentation
- Week 2 Perception and Bayesian Inference for Colour Segmentation
- Week 3 Video Processing and Motion Estimation
- Week 4 Deep learning and motion estimation
- Week 5 Applications in Video Processing
- Week 6 Another introduction to compression
- Week 8-12 Modern Video Compression and the Engineering behind YouTube and Netflix

Practical Work

- Week 1+2 : NUKE for Compositing
- Week 3 : Scripts in NUKE for video processing
- Week 4 – 6 : Writing your own script/graph for colour keying in video and writing a 4 page report
- Week 8 – 12 : Video Transcoding and Quality Measurement

Course Texts and Online Resources

Texts

- **Video Processing, Murat Tekalp**
- Markov Random Fields for Vision and Image Processing. Edited by A. Blake, P. Kohli and C. Rother, MIT Press, 2011. ISBN: 978-0-262-01577-6
- The Essential Guide to Video Processing. A. Bovik, Academic Press, 2009. ISBN: 978-0-12-374456-2

Online Resources

- IEEE Explore
- Various papers I will point out
- YouTube links as we go
- Shapiro's course on "Mars to Hollywood" and Katsagellos "Video and Image Processing" are references

The technical material discussed in this course is at the bleeding edge of what is possible. Get comfortable with not being able to find a complete set of notes like you'd find in a book which you can then just learn off. The book is still being written.

		9	10	11	12	13	14	15	16	17
Monday	Lecture	EEP55M08 - M17	EEMT17 - MMTLAB	EEMT17 - MMTLAB	EE5C01 M21	EEP55M08 - PBLT	EE5M02 - SPSR CS7NS6 - 1.07 LLOYD	EE5M02 - SPSR CS7NS6 - 1.07 LLOYD	EE5C01 - DO CS7NS5 - LB01	CS7NS5 - LB01 EE5M01 - M20
	Tutorial									
Tuesday	Lecture	EEP55M08 - SYNGE	EEP55C25 - CEDR CS7NS6 - 1.07 LLOYD	EEP55M08 - AP2.28 CS7GV5 - 4.09 LEINSTER		CS7GV3 - 1.20 LLOYD	CS7GV3 - 1.20 LLOYD EE5M02 - AP2.02	CS7DS2 - LB04 EE5C01 - AP2.28	EE5C01 - AP2.28	EE5M01 - AP2.28
	Tutorial									
	Lab									
Wednesday	Lecture	EEP55C24 - AP2.04	EEP55C24 - AP2.04	CS7GV5 - 4.09	EE5M02 - AP2.02		CS7NS2 - 1.07 LLOYD	CS7DS2 - LB08	ME5E3 - PBENG	ME5E3 - PBENG CS7NS5 - LB04
	Tutorial									
	Lab									
Thursday	Lecture		CS7IS2 - LB01		EEP55C24 - AP2.04	EEP55C24 - M17	EEP55C23 - PBENG	EEP55C23 - PBENG CS7CS3 - LB01	EEP55C25 - AP2.28 CS7CS3 - LB01	EEP55C25 - AP2.28 CS7CS3 - LB01
	Tutorial									
	Lab									
Friday	Lecture	EE5M01 AP2.28	EE5M01 - AP2.28	EE5C01 - AP2.28 CS7NS6 - 1.07 LLOYD	EE5C01 - AP2.28	CS7NS2 - 1.07 LLOYD	CS7IS2 - LB08			MECHLB
	Tutorial									
	Lab									

LECTURES/Demos

NUKE (weeks 1,2,3), Assignment time weeks 4-6
Compression Lab week 6, Assignment time weeks 9-12

ClassTests/Industry Talks

Some speakers in CA and NZ
so we'll have to get creative

		Week											
		1: Jan 24th	2 : Jan 31st	3: Feb 7th	4: Feb 14th	5: Feb 21st	6: Feb 28th	7: 7th Mar	8: 14 Mar	9: 21st Mar	10: 28 Mar	11: Apr 4	12: Apr 11
		Segmentation and Perception		Motion and Video		Denoising and Colour and Engineering Standards			Compression	Adaptive Streaming	Adaptive Streaming	DNNs for Compression	Exam Review
Monday 12 - 13:00	Lecture	Intro	Perception	Motion 1	Motion 2	Denoising	Motion		Intro and Compression	Cancel			
Monday 16 - 17:00	Lecture	Bayes	Perception	Motion	DNN	Motion	Motion		Compression	ABR DASH			
												Ioannis Talk	
Tuesday 15-17	Lecture	Keying	Keying	Bayes NUKE	MotionNUKE	NUKE	Perception		Compresson				
	Labs	NUKE	NUKE	Ass	Ass	Ass							
Friday 11-13:00	Lecture	Bayes	Colour		DNN	Denoising	Perception		RD Curves Lab	RD Curves Lab	ABR Bitrate Ladder Lab	ABR Bitrate Ladder Lab	ABR Bitrate Ladder Lab
	Labs	NUKE	NUKE	Ass	Ass	Ass							

Keynotes

From last
year

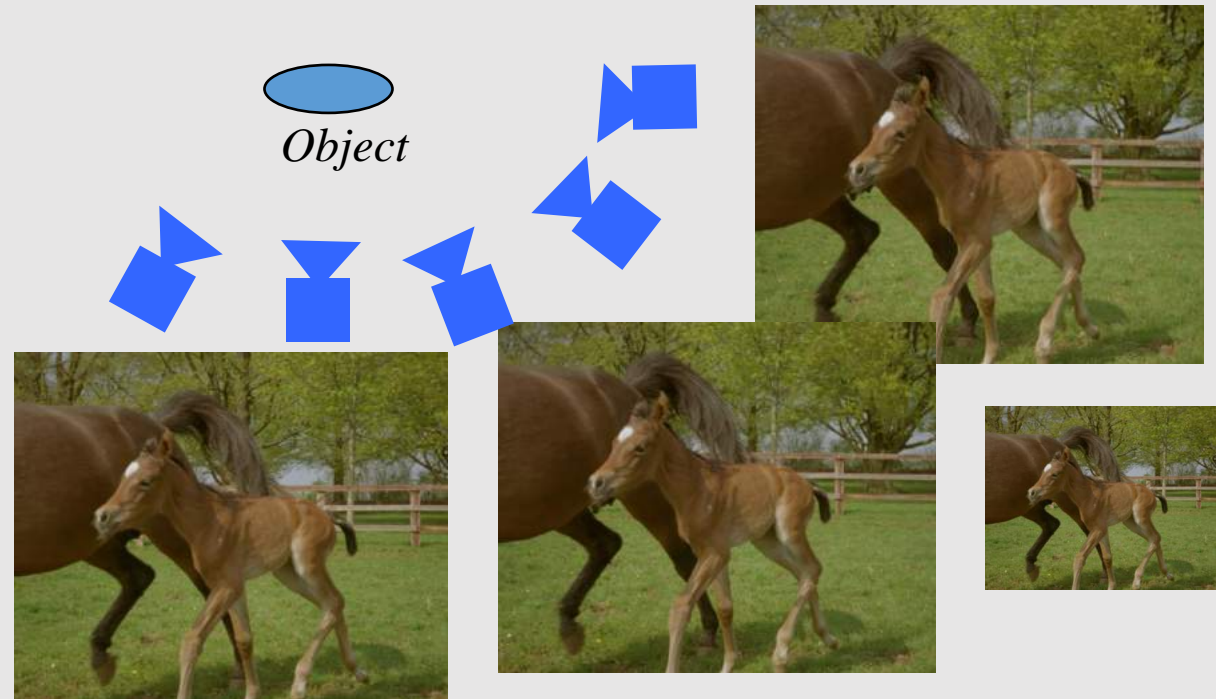
- [Simon Robinson](#) (Academy Award Winner from the Foundry, London)
- [Peter Hillman](#) (Academy Award winner from Weta Digital, New Zealand)
- [Peyman Milanfar](#) (R&D Lead for Pixel Cameraphone at Google, California)
- [Marta Mrak](#) (BBC R&D, London)

TBC

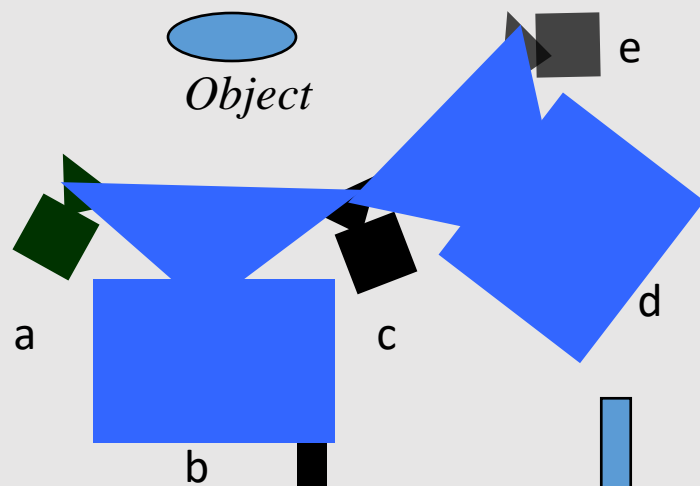
Ioannis Katsavounidis (Video Infra Lead at Facebook)

Paul Wilkins
(Google/Chrome)

Bullet Time







Cameras too big to be close enough!

A?

b

C?

d

E?



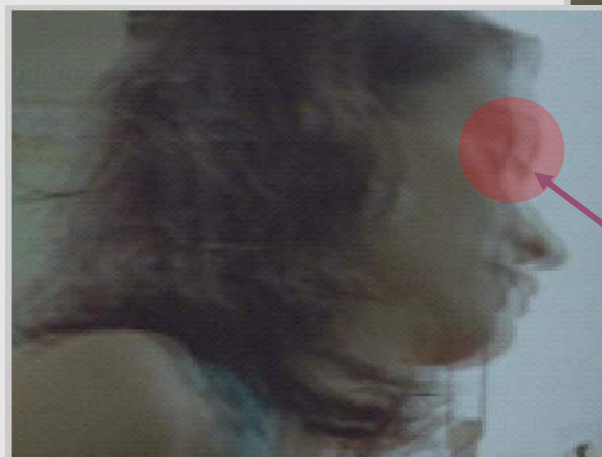
Thinking about “inbetweening”



Current



Next

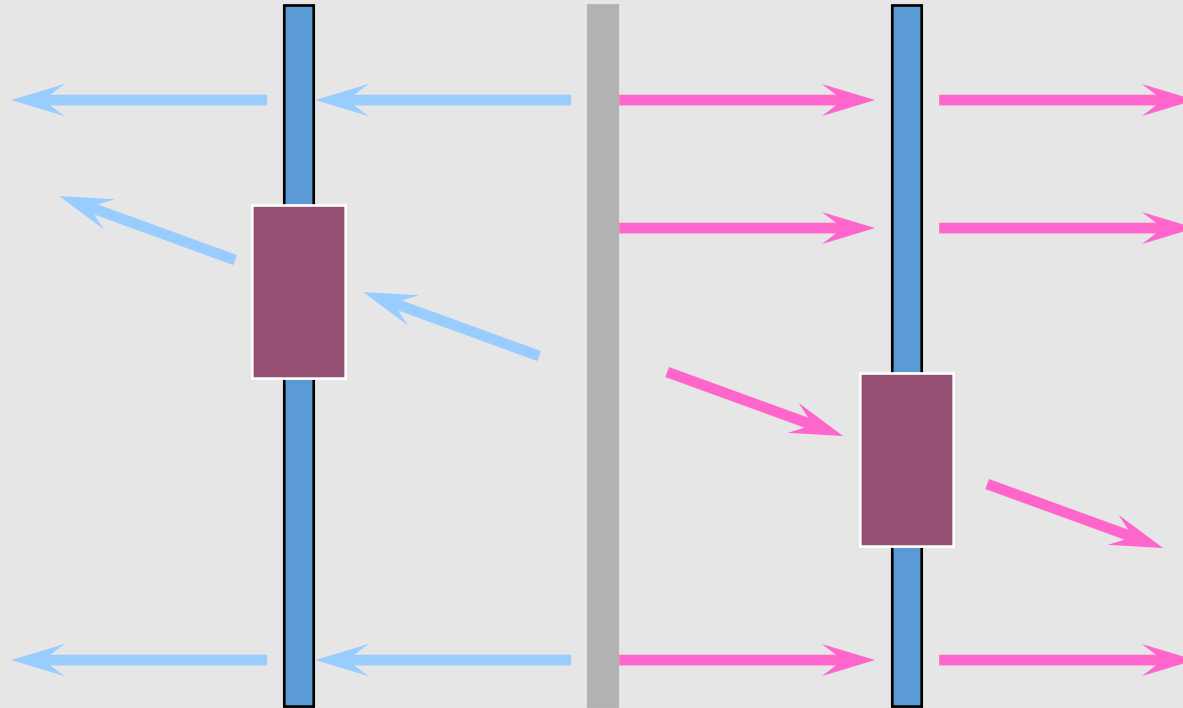


Blurring due to motion

$(\text{Current} + \text{Next}) / 2$

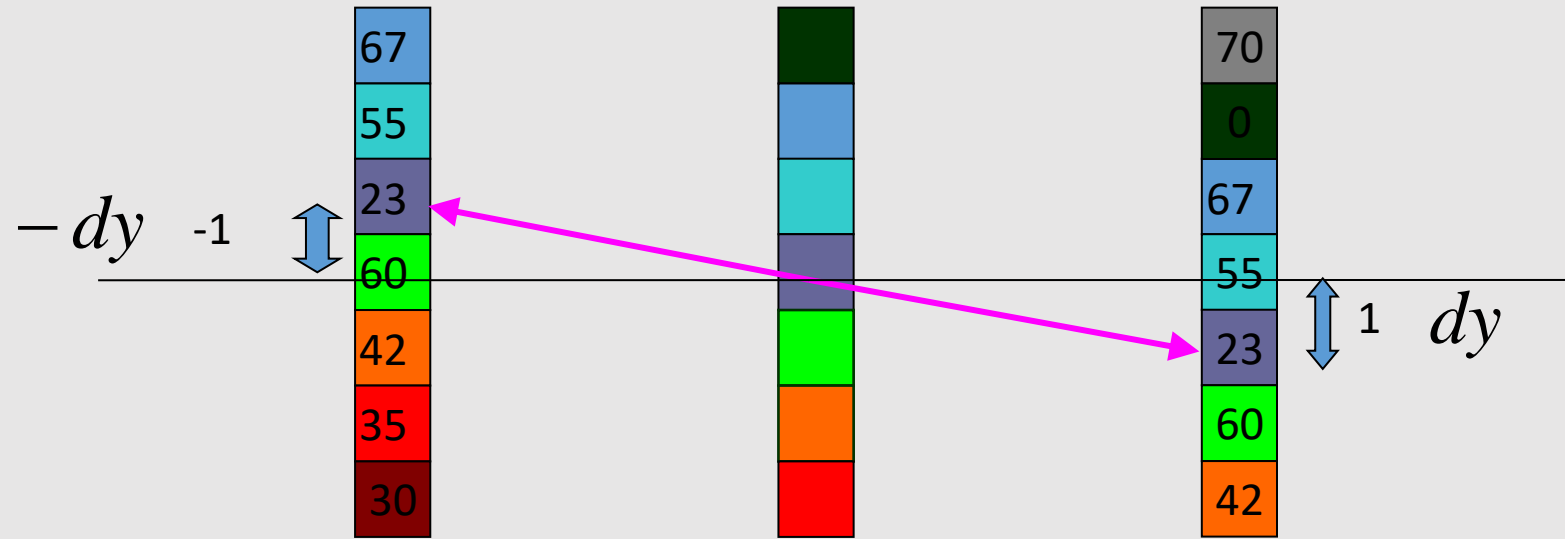
Some observations that help

The cameras always
cause smooth
movement



When things move
They don't change colour much

Bullet Time : Mathematical modelling

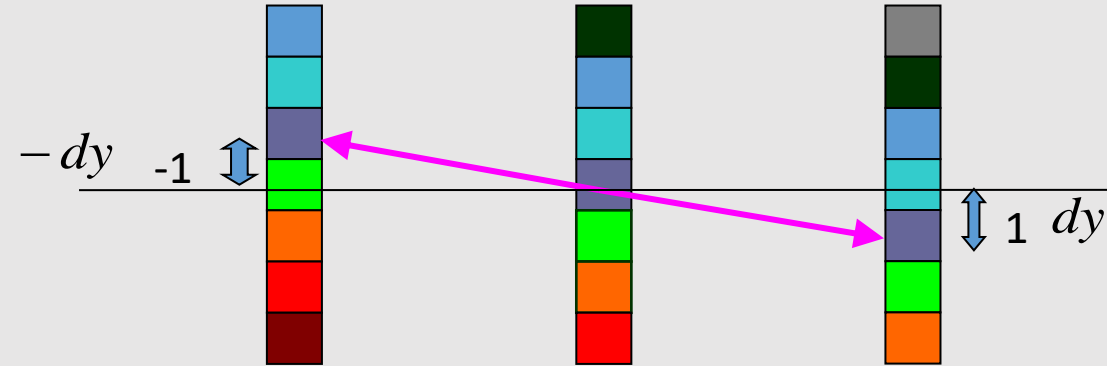


$$I_{n-1}(k - dy) = I_{n+1}(k + dy)$$

Bullet Time : How to solve for dy

$$I_{n-1}(k - dy) = I_{n+1}(k + dy)$$

$$\Rightarrow I_{n-1}(k - dy) - I_{n+1}(k + dy) = 0$$



Tricky because we have no analytic function for $I()$

So change this into an "optimisation problem"

Choose dy such that it makes $(I_{n-1}(k - dy) - I_{n+1}(k + dy))^2$ as small as possible

Minimise $(I_{n-1}(k - dy) - I_{n+1}(k + dy))^2$ wrt dy

Bullet Time : The Equation is actually 2D!

$$I_{n-1}(h-dx, k-dy) = I_{n+1}(h+dx, k+dy)$$

Minimise $(I_{n-1}(h-dx, k-dy) - I_{n+1}(h+dx, k+dy))^2$ wrt (dx, dy)

Oh crap.

There is no analytic solution. We don't have an explicit form for $I(x,y)$

And the solution is HUGE.

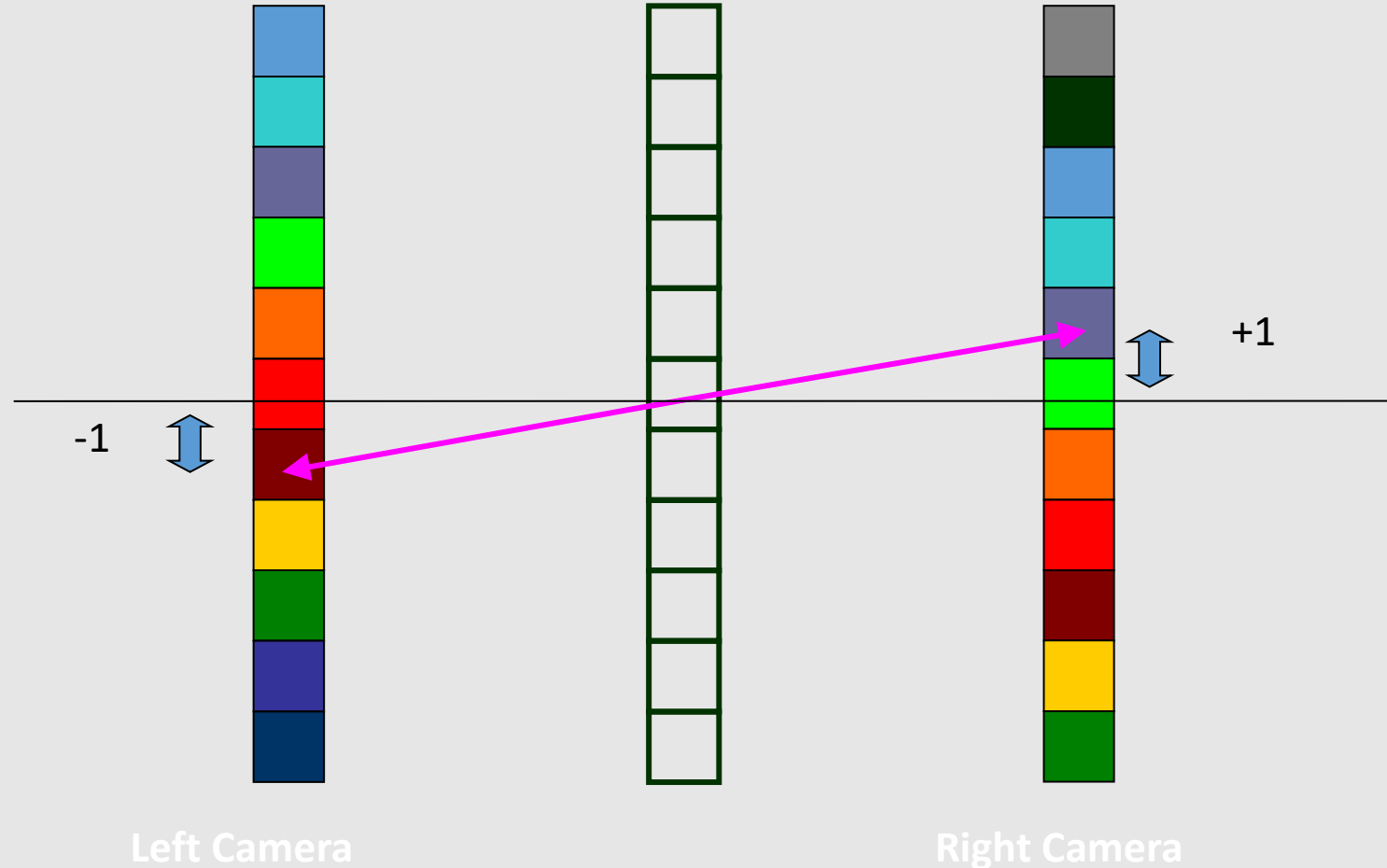
There is a dx and dy for EVERY PIXEL in EVERY frame!

4 Million Variables!

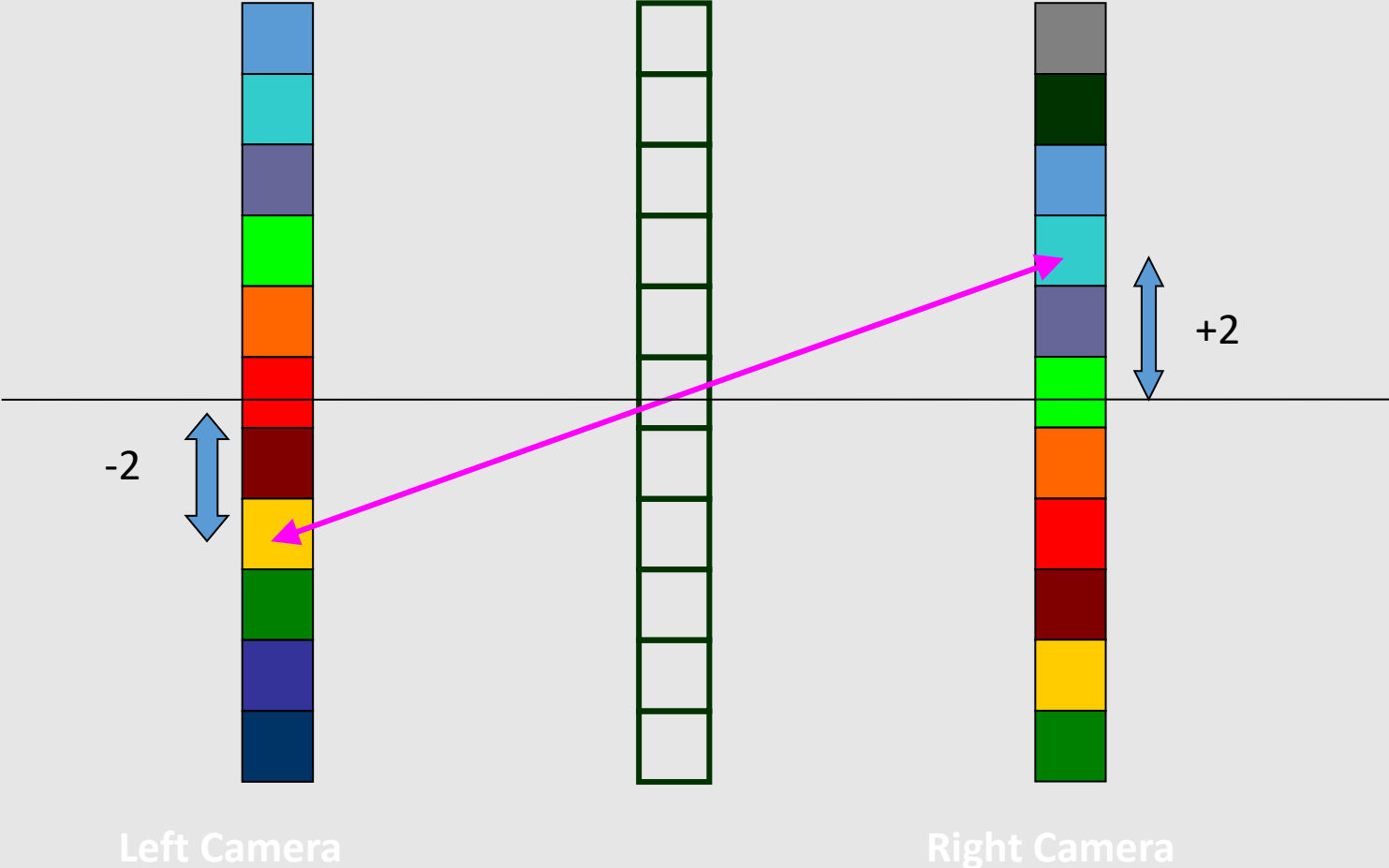
Brute force numerical solution : examine subset of possible pixel pairs and pick the pair which gives least “squared error”

$$E(dy) = (I_{n-1}(k - dy) - I_{n+1}(k + dy))^2$$

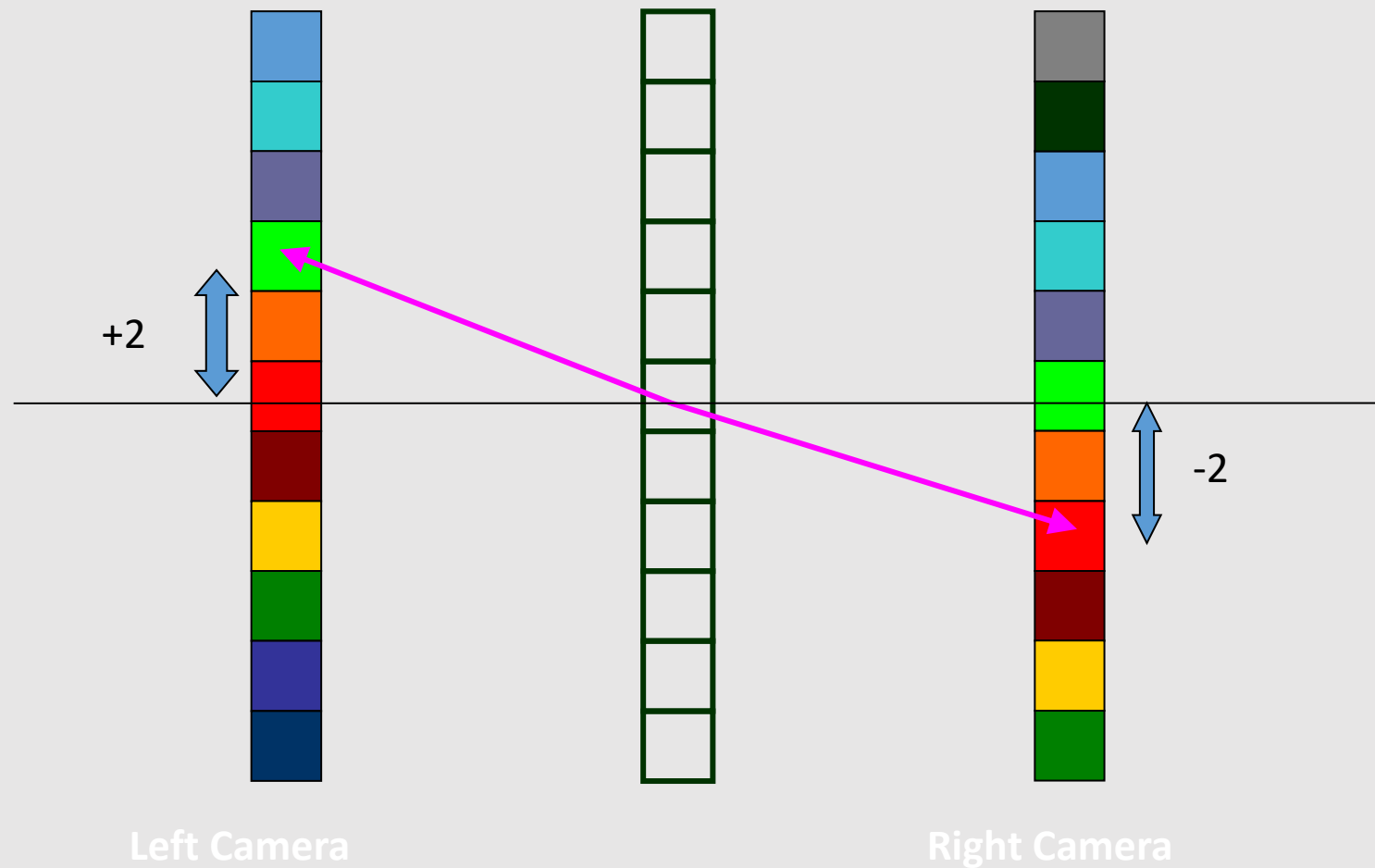
$$E(dy = 1) = 27$$



$E(dy = 1) = 27$
 $E(dy = 2) = 53.2$



$$E(dy = 1) = 27$$
$$E(dy = 2) = 53.2$$
$$E(dy = -2) = 10.2$$

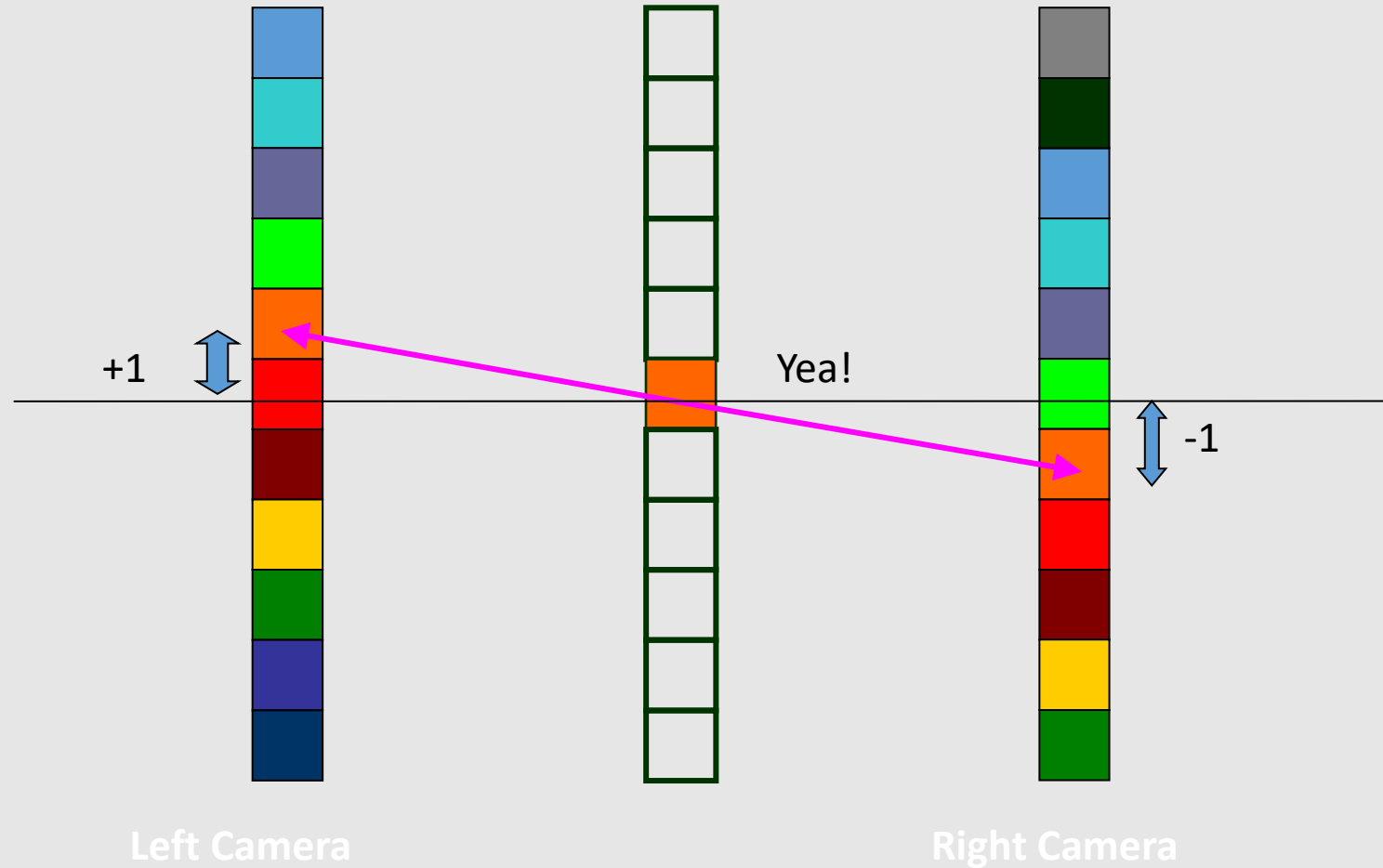


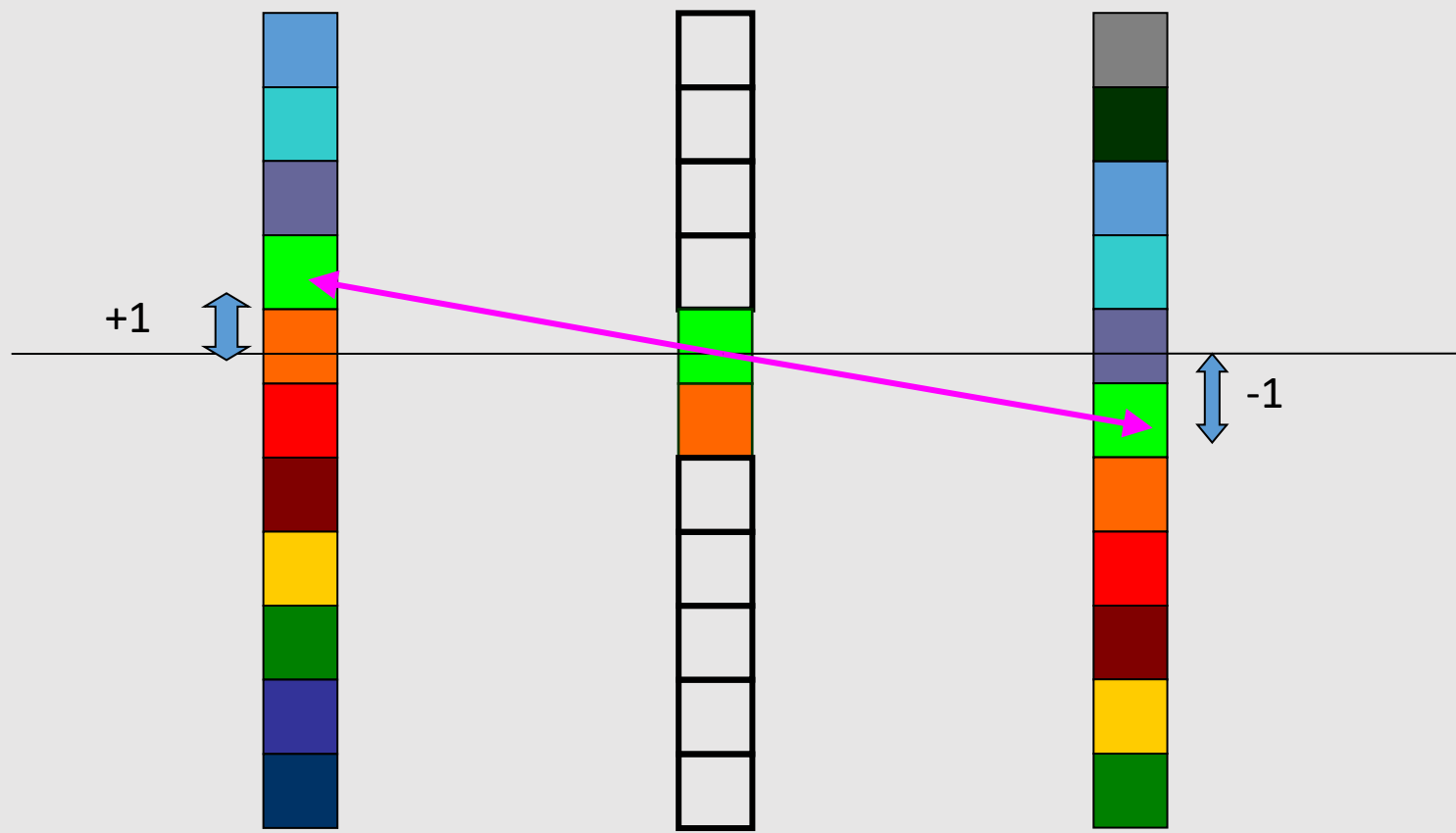
$$E(dy = 1) = 27$$

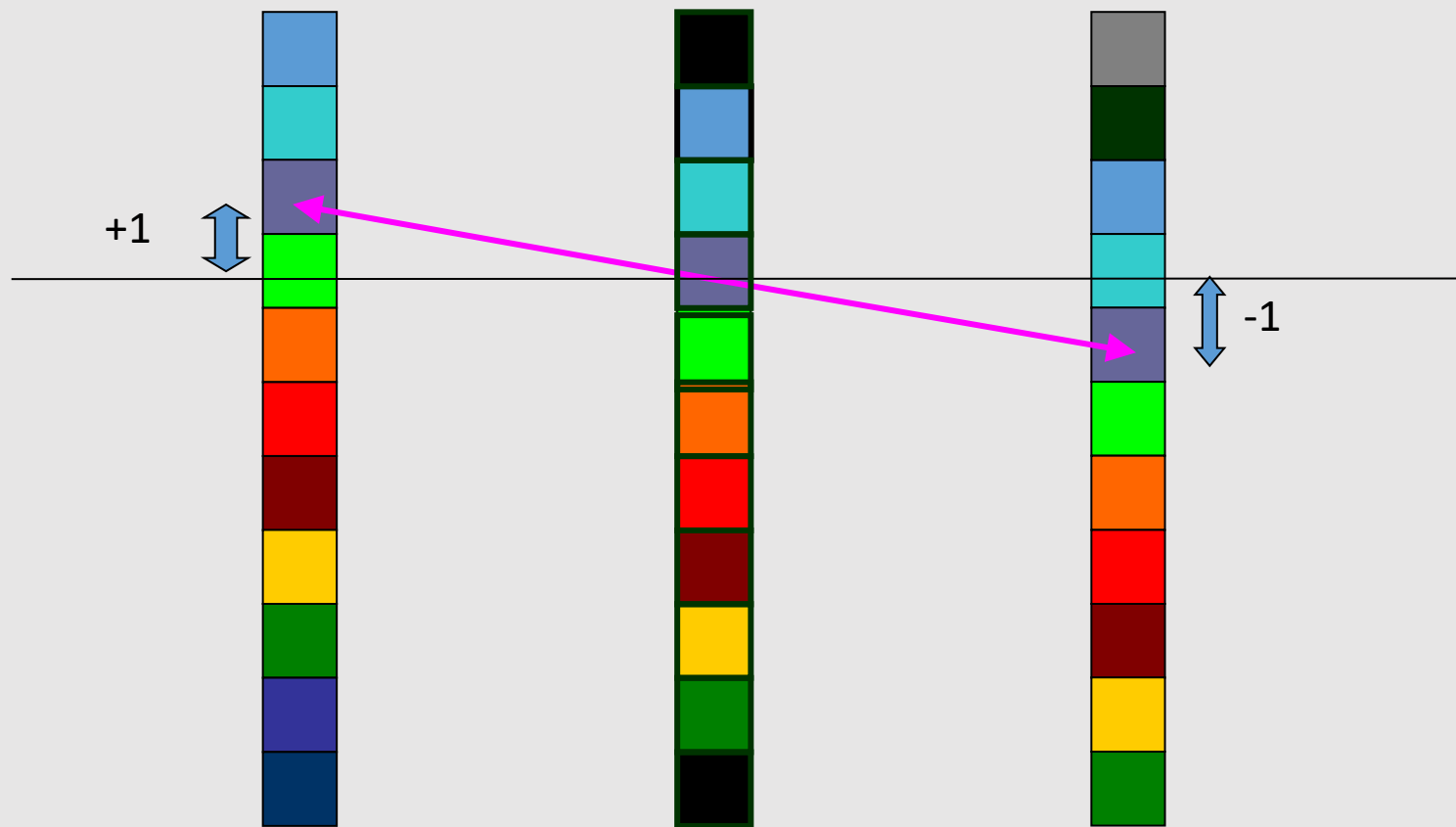
$$E(dy = 2) = 53.2$$

$$E(dy = -2) = 10.2$$

$$E(dy = -1) = 1.2$$







What about implementation?

```
function [new_frame] = inbetween(left, right, w)
    %Inputs are left and right vectors of pixels, and w the search space
    rows = length(left);
    new_frame = zeros(size(left));
    for k = w + 1 : rows - w,
        error = zeros(w*2+1,1);
        %At this pixel location "k", calculate the error
        %associated with every possible offset dy
        for dy = -w : w,
            error(dy + w + 1) = (left(k - dy) - right(k + dy))^2;
        end;
        %Select the dy that minimised the error
        [y, index] = min(error);
        estimated_dy = index - w - 1;
        %Use it to calculate an interpolated value
        new_frame(k) = (left(k - estimated_dy) + right(k + estimated_dy))/2;
    end;
```


Actually this simple “algorithm” has issues

- Search space is “integer” but the motion is “fractional”
- We need to work in colour i.e. 3 pixels at each location not one
- There is noise in real signals .. So the pixel differences will be “noisy”
 - That leads to random motion estimates
 - Which leads to “weirdly distorted” in-betweens.
- No guarantee that the motion estimates are “smooth enough”
- In real life sometimes you can’t find a match because objects move behind or in front of each other. So the “mathematical model” is not right everywhere.
 - That leads to objects sticking to background
- It is too dirt slow!

So we need better modelling etc etc

$$\begin{aligned}
 p &= \left(\frac{(g_n - i_n)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n-1})^2}{2\sigma_c^2} \right) p_s(s = [0, 0, 1] | S) p_c(c | C) \exp - \alpha & s = [0, 0, 1] \\
 p &= \left(\frac{(g_n - i_n)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n+1})^2}{2\sigma_c^2} \right) p_s(s = [0, 1, 0] | S) p_c(c | C) \exp - \alpha & s = [0, 1, 0] \\
 p &= \left(\frac{(g_n - i_n)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n-1})^2 + (i_n - i_{n+1})^2}{2\sigma_c^2} \right) p_s(s = [0, 0, 0] | S) p_c(c | C) & s = [0, 0, 0] \\
 p &= \left(\frac{(g_n - c)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n-1})^2 + (i_n - i_{n+1})^2}{2\sigma_c^2} \right) p_s(s = [1, 0, 0] | S) p_c(c | C) & s = [1, 0, 0] \\
 p &= \left(\frac{(g_n - c)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n-1})^2}{2\sigma_c^2} \right) p_s(s = [1, 0, 1] | S) p_c(c | C) \exp - \alpha & s = [1, 0, 1] \\
 p &= \left(\frac{(g_n - c)^2}{2\sigma_\mu^2} \right) \exp - \left(\frac{(i_n - i_{n+1})^2}{2\sigma_c^2} \right) p_s(s = [1, 1, 0] | S) p_c(c | C) \exp - \alpha & s = [1, 1, 0]
 \end{aligned}$$

distributions for the random variables $\mathcal{N}(x|x_1, \sigma_1^2)$ and $\mathcal{N}(x|x_2, \sigma_2^2)$; with means and variances x_1, σ_1^2 and x_2, σ_2^2 it can be shown that

$$\mathcal{N}(x|x_1, \sigma_1^2) \mathcal{N}(x|x_2, \sigma_2^2) dx = \frac{1}{2\pi \sqrt{\sigma_1^2 \sigma_2^2}} \exp - \left[\frac{(\bar{x} - x_1)^2}{2\sigma_1^2} + \frac{(\bar{x} - x_2)^2}{2\sigma_2^2} \right]$$

$(\sigma_1^2 + \sigma_2^2)$. See [70] for some background material on integrating Gaussians.





x4 Repeat Frames



x4 MotionReTime

Goro@Welsh corgi

