graphics and vision gravis

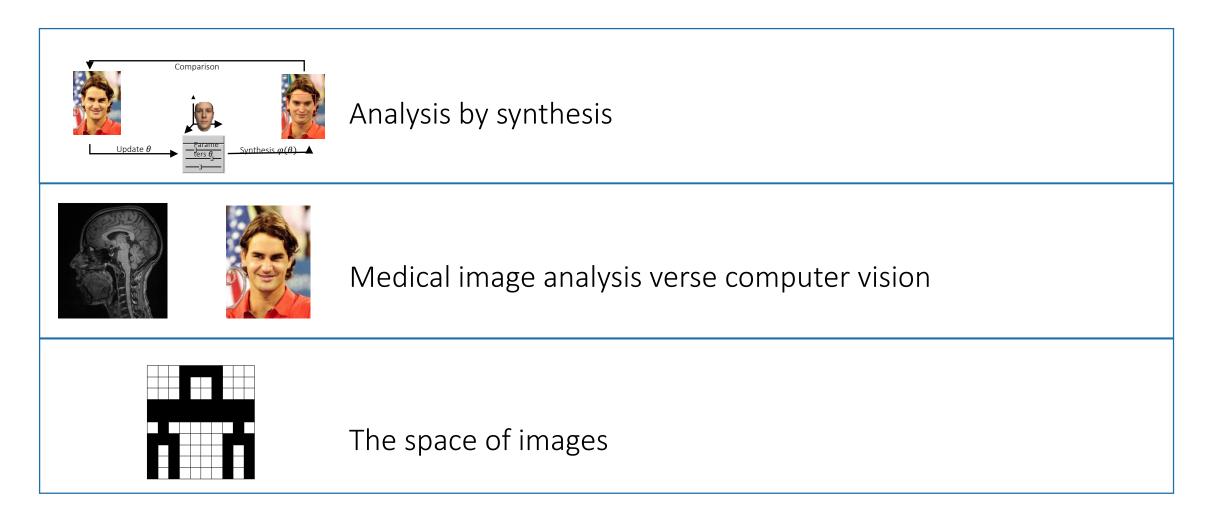


Probabilistic Morphable Models - An overview -

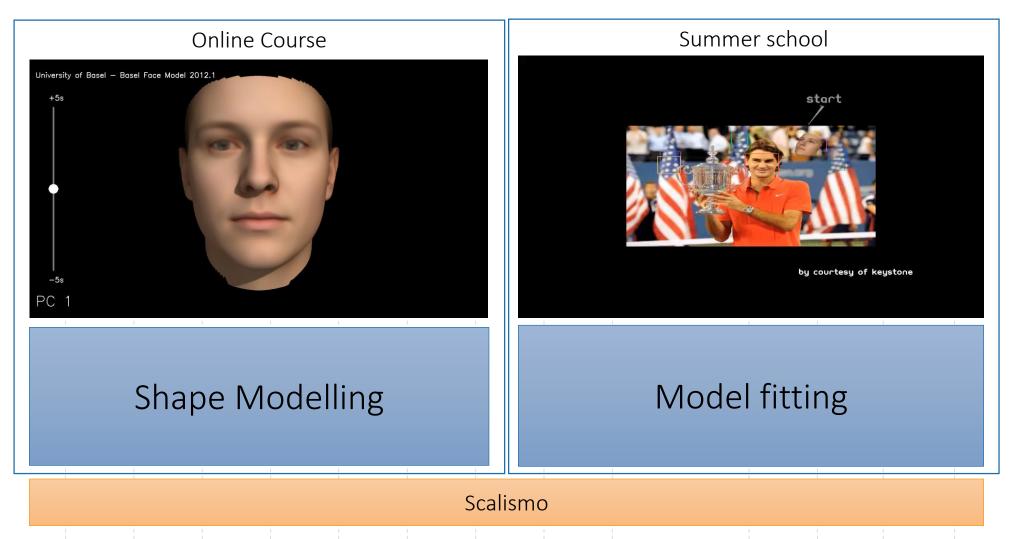
Marcel Lüthi

Graphics and Vision Research Group Department of Mathematics and Computer Science University of Basel

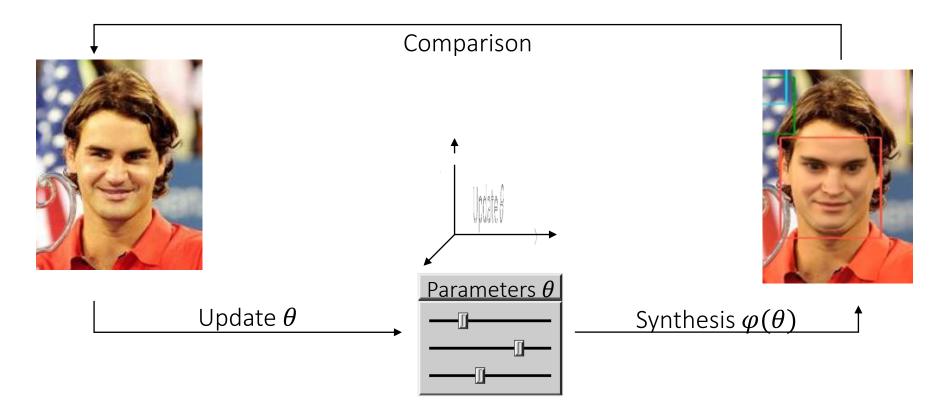
Outline



Probabilistic Morphable Models

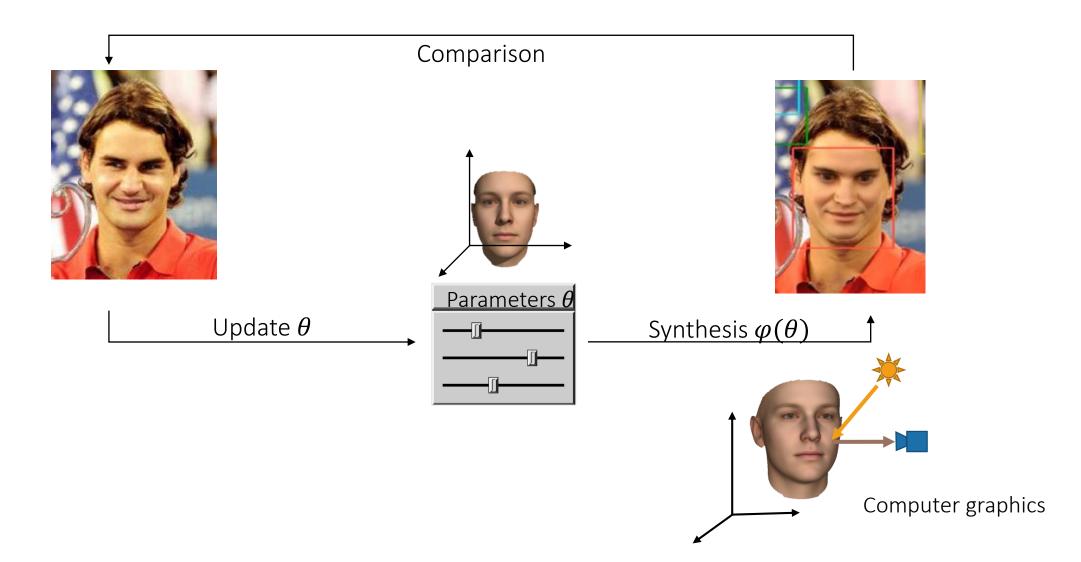


Conceptual Basis: Analysis by synthesis

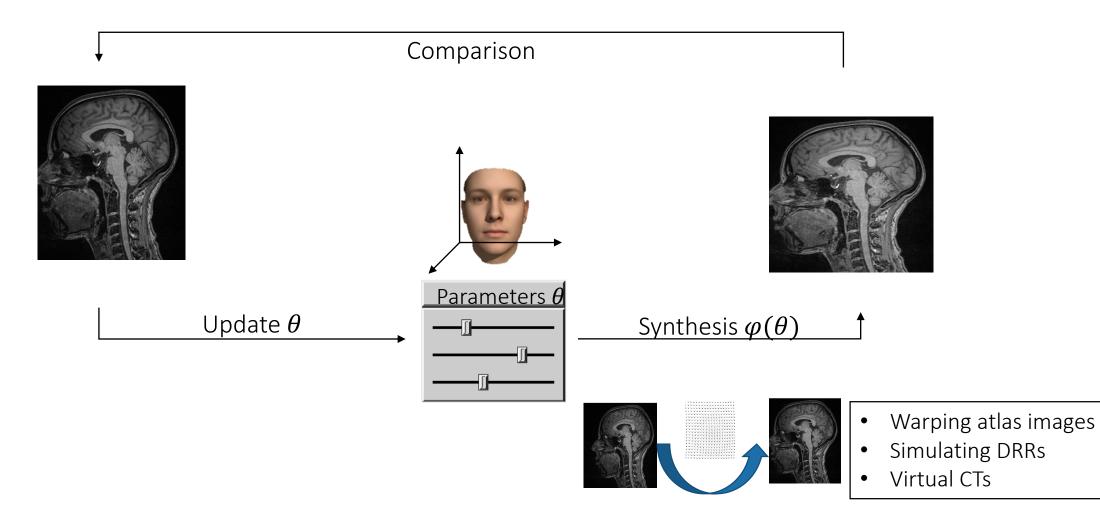


- If we are able to synthesize an image, we can explain it.
 - We can explain unseen parts and reason about them

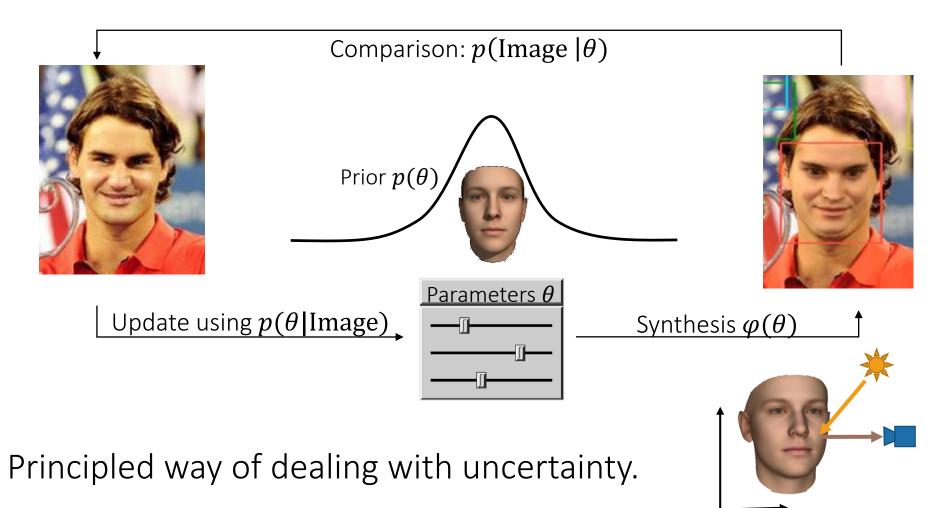
Synthesizing images



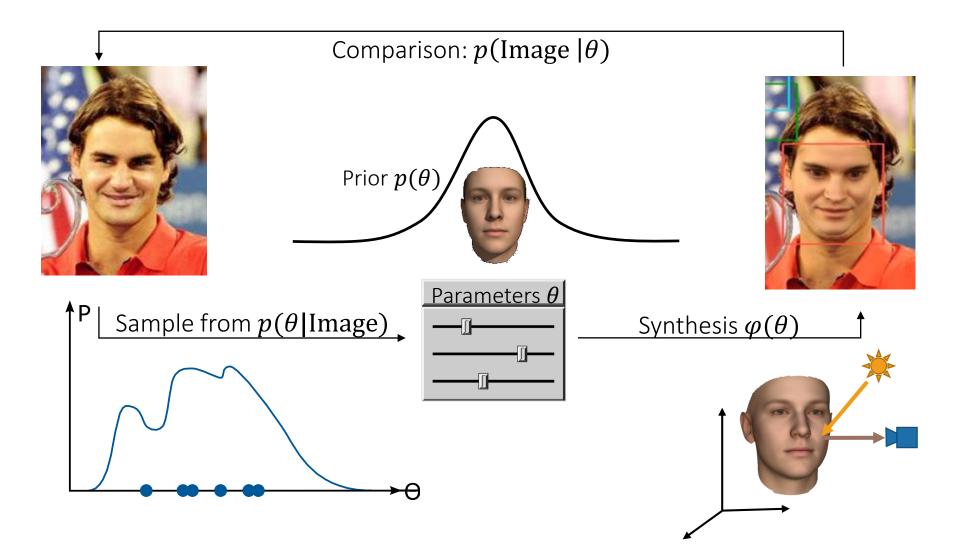
Synthesizing images



Mathematical Framework: Bayesian inference



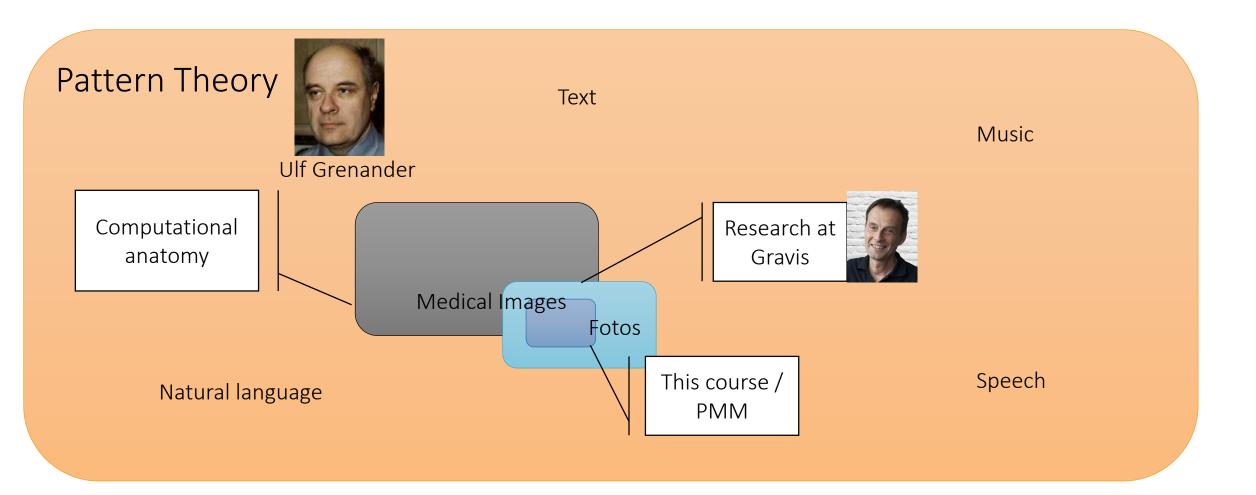
Algorithmic implementation: MCMC



Course programme

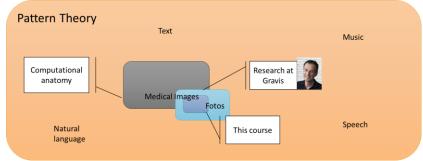
	Morning	Afternoon
Tuesday	 Introduction The course at a glance Basics of Computer Graphics 	 Basic tasks in Scalismo-faces Bayesian modelling Welcome reception
Wednesday	 Probabilistic model fitting using MCMC 	Exercises: MCMC FittingIntroduction to course project
Thursday	Face image analysis	Course project
Friday	 Connections to medical image analysis Advanced topics in Gaussian processes 	Course project
Saturday	Project presentation	Social event

The course in context

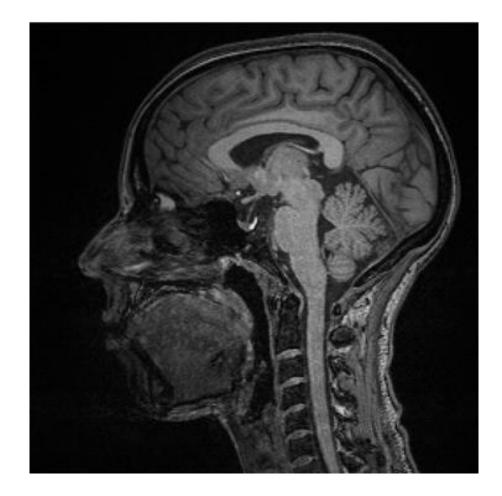


Pattern theory vs PMM

- Pattern theory is about developing a theory for understanding realworld signals
- Probabilistic Morphable Models are about **using** theoretical well founded concepts to analyse images.
 - GPs for modelling
 - MCMC for model fitting
 - Working software



Images: Medical Image Analysis vs Computer Vision





Source: OneYoungWorld.com

Images in medical image analysis

Goal: Measure and visualize the unseen

- Acquired with specific purpose
 - Controlled measurement
 - Done by experts
- Calibrated, specialized devices

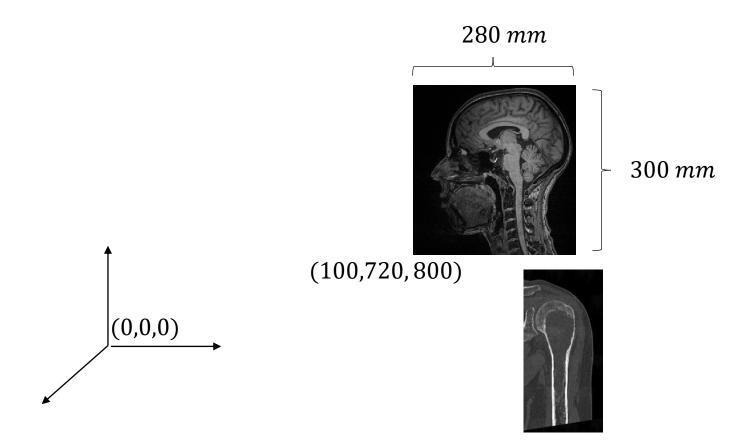


Images in medical image analysis



• Images live in a coordinate system (units: mm)

Images in medical image analysis

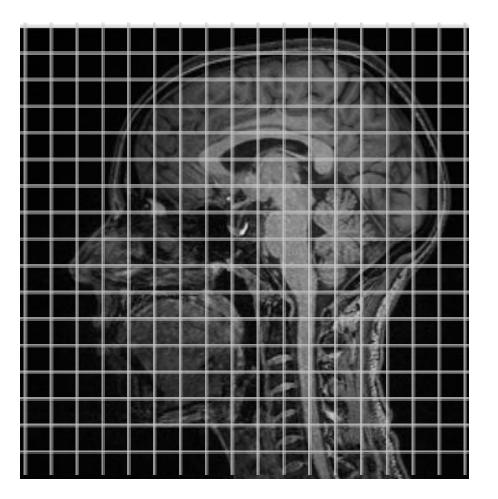


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Images in medical image analysis

Values measure properties of the patient's tissue

- Usually scalar-valued
- Often calibrated
- CT Example: -1000 HU -> Air 3000 HU -> cortical bone



Images in computer vision

Goal: Capture what we see in a realistic way

- Perspective projection from 3D object to 2D image
 - Many parts are occluded



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Images in computer vision

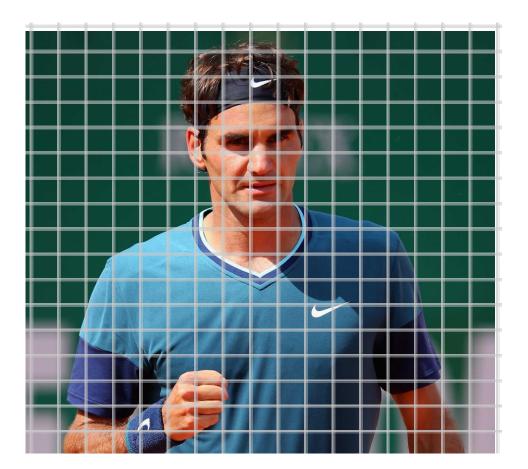
- Can be done by anybody
 - Acquisition device usually unknown
 - Uncontrolled background, lighting, ...
- No clear scale
 - What is the camera distance?
- No natural coordinate system
 - Unit usually pixel



Source: twitter.com

Images in computer vision

- Pixels represent RGB values
- Values are measurement of light
 - Reproduce what the human eye would see
- Exact RGB value depends strongly on lighting conditions
 - Shadows
 - Ambient vs diffuse light



Images: Medical Image analysis vs Computer Vision

Medical image

- Controlled measurement
- Values have (often) clear interpretation
- Explicit setup to visualize unseen
- Coordinate system with clear scale

Computer vision

- Uncontrolled snapshot
- Values are mixture of different (unknown factors)
- Many occlusion due to perspective
- Scale unknown

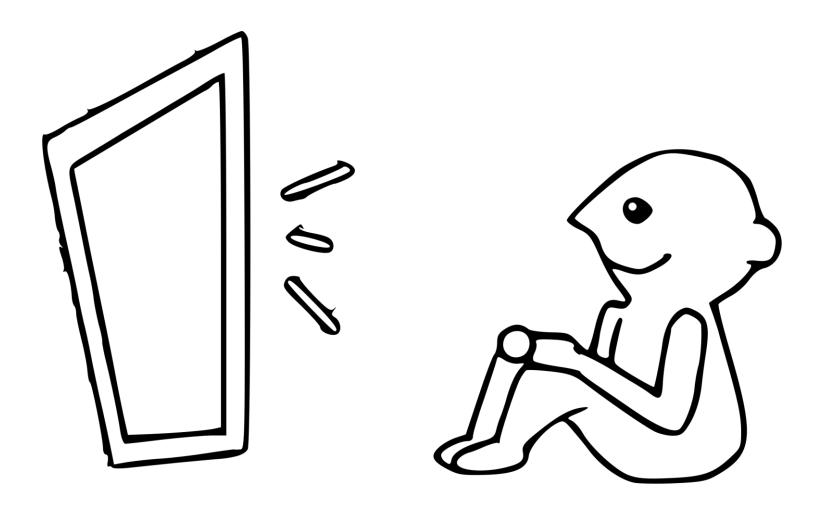
Many complications of computer vision arise in different form also in a medical setting.

> DEPARTMENT OF MATHEMATICS AND COMPUTER SCIENCE

The space of images

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The space of images

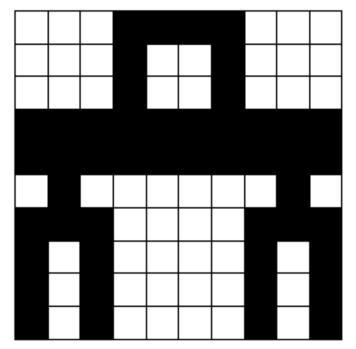


The space of images

- 10 x 10 image
- 2 colours

How long would you need to watch TV (24 fps) until you have seen all such images?

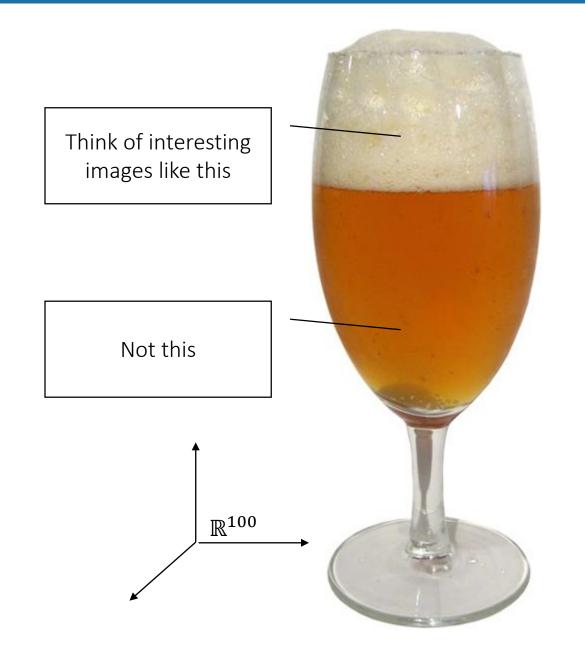
- Number of images: $2^{100} \approx 1.2e30$
- Watching 100 years continuously
 24 x 60 x 60 x 24 x 365 x 100 ≈7.5e10



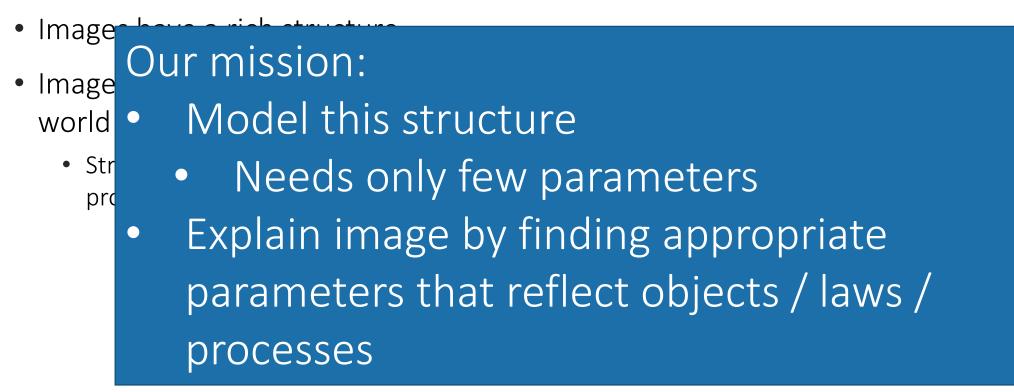
Source: bbc.co.uk

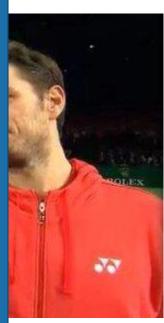
The space of images

- Most images are uninteresting
- Only very few of all possible images are of interest to us



Structure in images





Next stop: Computer graphics

