

## Computer Vision:

# Geometry, uncertainty and deep learning

#### Roberto Cipolla Department of Engineering

http://www.eng.cam.ac.uk/~cipolla/people.html http://www.toshiba.eu/eu/Cambridge-Research-Laboratory/

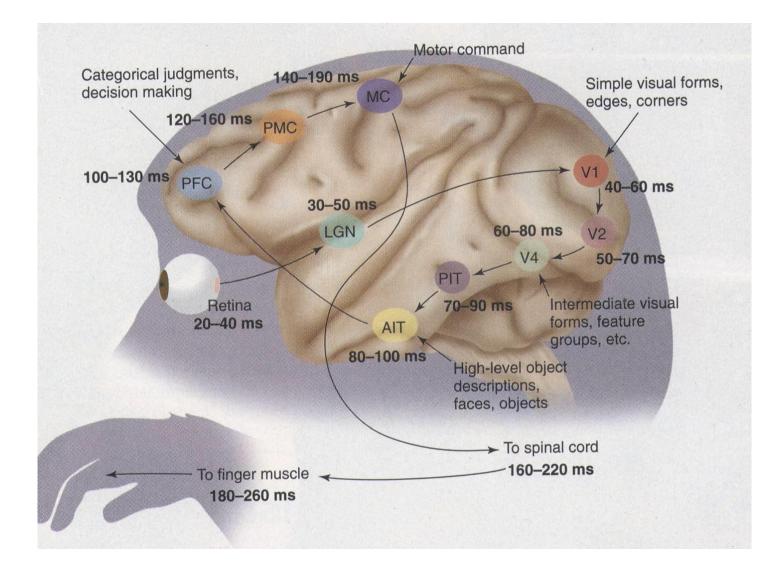


"Intelligence can be viewed as a process that converts unstructured information into useful and actionable knowledge" - Hassabis (DeepMind)



#### Vision: what is where by looking





#### **Computer Vision**





#### **Computer Vision – What?**



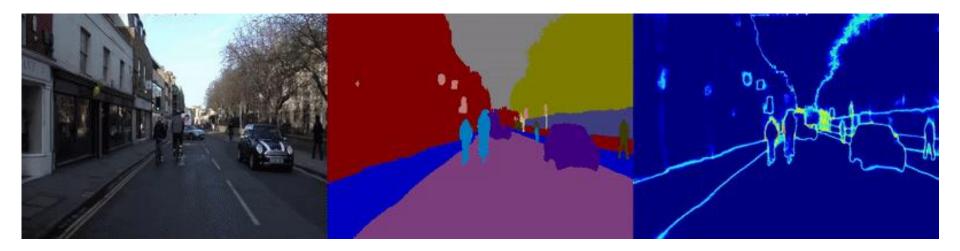
158 237 087 251 255 249 253 042 251 255 197 253 247 165 157 250 255 255 255 255 255 188 120 090 123 140 146 255 244 028 024 254 249 049 039 245 183 140 085 023 029 255 165 019 255 180 212 197 236 245 109 066 190 251 255 252 088 253 255 144 125 075 245 255 256 256 256 256 256 266 194 045 092 181 096 245 086 211 234 031 016 180 255 047 029 188 236 254 120 041 032 226 193 043 203 246 255 255 158 254 059 069 255 255 251 237 114 248 098 254 241 254 255 250 215 253 123 097 255 032 031 132 153 038 137 009 032 028 185 075 039 031 115 247 100 089 140 034 177 160 059 088 247 245 255 200 080 075 156 221 169 247 222 185 039 142 247 255 248 255 246 255 255 255 255 255 040 255 080 066 233 039 142 250 029 023 253 036 033 036 117 234 245 112 220 036 255 058 167 051 044 167 043 145 085 082 200 035 239 225 255 246 028 042 248 255 253 065 253 255 239 247 186 151 058 252 155 137 184 255 210 255 043 025 255 130 031 041 137 255 163 189 225 022 223 031 041 107 033 086 238 220 133 117 248 153 179 211 250 255 156 195 255 247 245 062 245 250 255 255 199 229 021 021 253 255 242 037 082 255 053 021 255 025 033 033 112 216 021 255 255 033 076 083 080 111 229 052 126 244 223 157 142 207 168 224 253 255 157 255 191 246 249 114 249 251 255 255 255 236 076 031 030 255 218 198 192 255 057 022 244 022 032 032 052 222 028 219 253 035 157 159 027 061 202 102 098 167 180 198 142 227 247 075 065 135 075 114 168 021 128 226 254 255 255 255 255 255 255 255 255 165 255 152 045 017 244 040 031 031 165 131 079 255 072 031 033 022 127 062 253 171 235 071 253 236 124 255 063 083 084 059 032 069 054 087 094 147 069 230 255 249 255 249 255 230 255 190 177 219 185 116 210 022 255 037 025 015 255 025 230 255 247 029 149 020 046 069 192 239 108 050 038 040 036 037 084 149 052 046 037 052 045 108 126 134 250 064 055 051 050 056 070 038 045 024 042 084 097 022 058 021 019 032 029 015 035 023 071 224 255 033 211 052 058 060 070 255 230 044 055 046 032 042 091 170 084 095 035 062 049 097 135 137 056 052 034 047 041 053 037 029 026 043 025 037 044 045 055 017 017 028 026 032 042 015 020 084 231 031 021 030 029 076 145 213 254 054 059 049 019 044 054 052 084 035 035 034 042 057 063 106 065 047 117 044 072 067 064 050 084 062 028 058 058 058 058 022 035 027 023 043 041 015 035 034 046 025 035 026 034 035 254 116 199 020 070 048 024 030 051 084 041 076 032 042 074 007 082 089 083 088 088 088 088 042 029 053 060 020 061 070 049 043 027 032 038 035 041 044 034 034 032 038 035 044 039 145 226 237 050 067 040 029 039 027 052 065 092 035 060 101 117 092 086 063 071 076 072 077 030 060 025 059 047 030 040 047 058 053 032 036 027 031 053 049 028 034 043 036 027 041 057 033 035 235 138 253 055 065 050 030 081 032 057 068 068 033 060 079 098 088 084 087 044 036 068 055 032 061 045 061 024 039 045 043 088 031 035 029 035 040 034 057 030 055 024 027 026 036 035 025 198 242 025 055 071 086 023 022 039 069 068 029 036 055 083 103 095 083 079 078 051 080 088 080 057 031 061 048 023 049 101 105 084 057 035 024 050 041 040 035 022 041 024 025 030 030 027 032 079 179 255 053 074 049 030 042 049 076 088 037 031 051 076 084 125 077 098 094 068 069 048 085 040 057 057 067 016 046 050 048 072 056 031 020 037 055 040 045 023 027 020 033 029 047 031 045 147 237 255 048 057 032 026 042 040 060 080 064 030 050 076 089 081 068 073 087 064 055 055 083 032 067 063 084 022 044 029 043 083 073 030 029 054 026 025 045 023 024 022 018 031 037 028 036 151 255 255 056 052 067 026 093 051 081 065 029 033 039 049 095 128 052 090 034 046 045 049 059 022 055 028 034 023 036 032 042 084 038 026 028 038 037 050 039 028 022 023 025 021 025 031 068 081 255 255 047 055 058 040 051 045 045 045 045 048 038 041 067 078 039 061 056 051 056 050 042 050 042 050 039 035 035 050 038 028 025 035 037 051 035 039 020 054 019 018 024 025 086 054 251 083 053 055 053 027 054 050 047 042 041 033 038 038 044 045 046 047 052 051 041 057 044 024 047 041 037 024 035 044 033 081 038 026 021 043 027 026 023 047 025 015 026 023 062 101 107 077 088 097 089 086 085 080 082 088 085 083 081 086 085 094 096 102 106 096 114 098 057 061 052 048 036 028 035 029 027 081 036 020 026 031 024 039 032 034 025 043 020 026 021 010 047 018 022 066 101 104 097 100 104 092 093 094 093 088 037 079 078 078 079 078 079 085 076 080 082 081 101 099 083 098 094 034 043 028 074 051 015 029 104 102 086 022 040 025 028 022 027 080 092 069 048 127 135 039 094 068 079 080 086 089 101 097 094 108 096 095 097 090 086 102 097 094 099 077 093 096 091 102 068 056 039 036 088 029 018 023 078 066 070 026 042 024 073 018 029 112 123 107 145 131 059 D70 D80 D83 D70 D88 D86 D77 D78 D81 D81 D80 D82 D74 D75 D73 D77 D73 D77 D73 D72 D90 D72 D16 D96 D91 D81 D81 D32 D34 D39 D90 D72 D12 D24 D97 115 143 D34 D49 D28 D01 D84 D77 129 136 118 113 D87 D35 054 085 118 077 082 066 071 078 081 061 018 059 055 044 049 086 056 054 051 058 015 065 078 077 082 073 027 037 034 106 016 015 027 081 109 095 031 046 060 085 067 140 121 113 091 120 085 012 052 088 060 062 048 063 065 040 031 013 022 045 023 054 094 030 048 032 040 017 062 082 071 068 068 023 031 032 112 028 021 025 118 033 057 068 102 117 117 101 093 090 082 109 062 067 059 051 080 054 045 016 048 064 037 041 043 016 038 056 036 061 094 021 041 032 040 017 036 055 059 058 082 014 028 032 120 071 053 071 050 074 045 113 124 131 132 121 105 124 124 082 085 086 085 043 078 022 036 038 047 071 045 039 043 021 040 041 046 044 099 134 047 043 015 110 038 059 047 075 080 021 071 069 071 075 106 102 110 127 130 131 129 120 124 119 111 072 083 080 090 093 090 061 072 032 035 027 024 061 028 062 056 027 042 034 020 050 083 040 026 052 037 035 036 040 071 041 070 084 082 100 117 106 105 095 115 113 110 110 108 096 082 102 127 081 086 092 093 089 087 025 077 033 028 019 023 057 035 016 050 025 058 025 058 025 058 025 058 080 087 077 094 102 102 102 115 098 102 099 095 086 087 101 099 095 103 090 092 113 084 059 095 089 086 088 091 091 088 085 047 071 042 051 053 062 070 085 090 075 083 093 099 099 103 093 104 084 069 073 092 076 080 090 084 093 103 095 087 095 088 089 075 111 084 091 096 079 099 086 095 091 088 086 076 077 073 077 088 090 094 070 100 084 072 078 081 099 106 101 096 087 086 082 089 085 066 067 083 079 081 073 088 092 085 074 068 049 070 065 055 060 091 112 063 082 073 076 075 083 088 095 091 094 095 095 072 077 075 077 066 067 068 065 062 061 060 066 059 061 060 065 061 073 069 065 071 050 045 045 039 038 029 038 053 074 141 055 068 078 102 096 097 095 095 097 095 094 099 095 097





#### **Real-time application - SegNet**



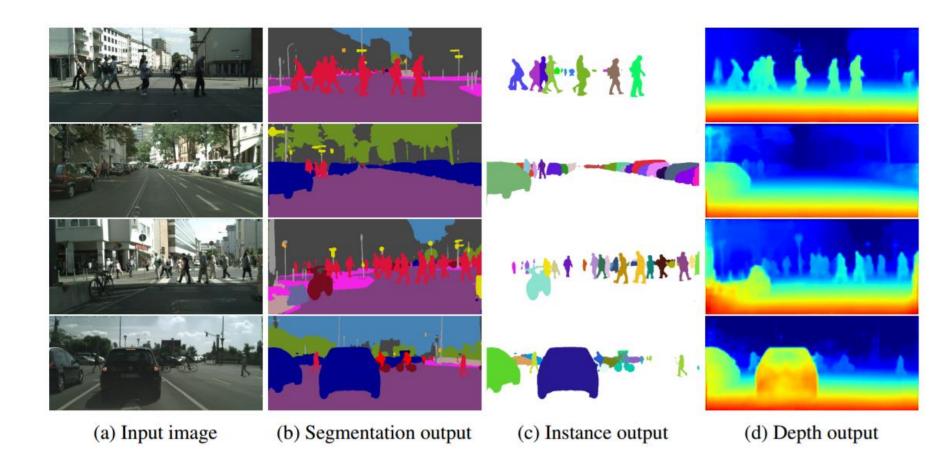


Input Image Semantic Segmentation Uncertainty

#### Badrinarayanan, Kendall and Cipolla 2015 and 2017 SegNet: Encoder-decoder architectures for scene segmentation

#### Multi-Task Learning

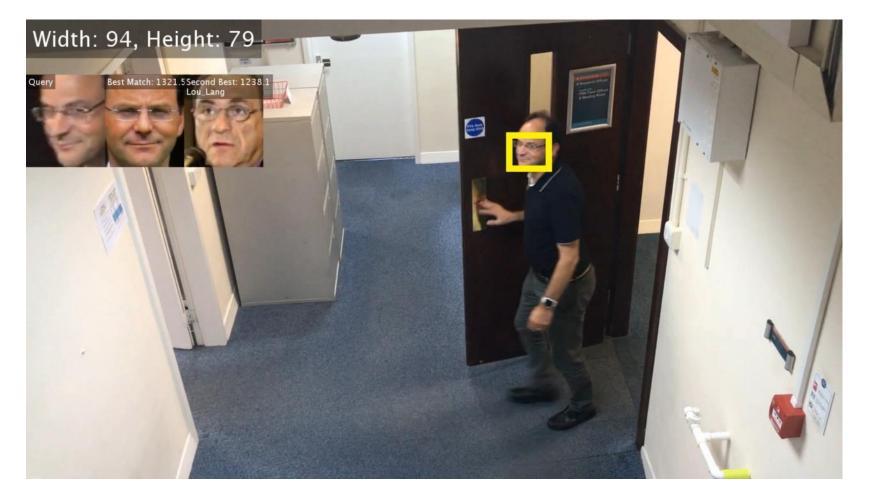




Kendall, Gal and Cipolla 2018 Multi-task Deep Learning

#### Deep learning – visual surveillance

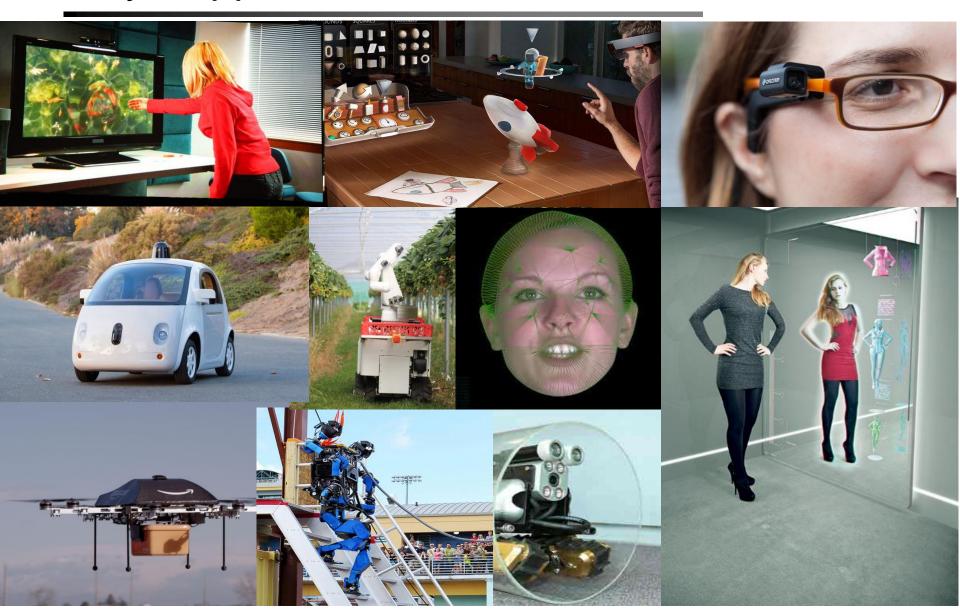




Cao et al. 2017 Real-time multi-person 2D pose estimation Charles et al. 2017 Real-time factored convnets (31 body parts)

#### Why? Applications





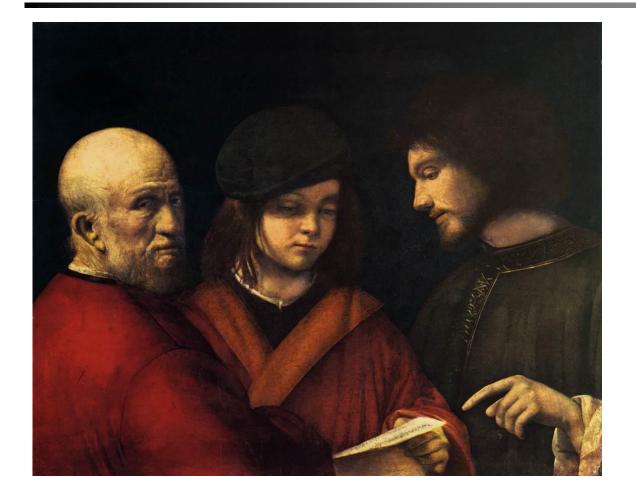
#### How?





#### Three faces of computer vision





#### The three ages of man - Giorgione



Geometrical framework Statistical framework Machine learning and data

- = reconstruction
- = registration
- = recognition





#### 1. Introduction

#### 2. 3R's of Computer Vision:

- Reconstruction
- Registration
- Recognition

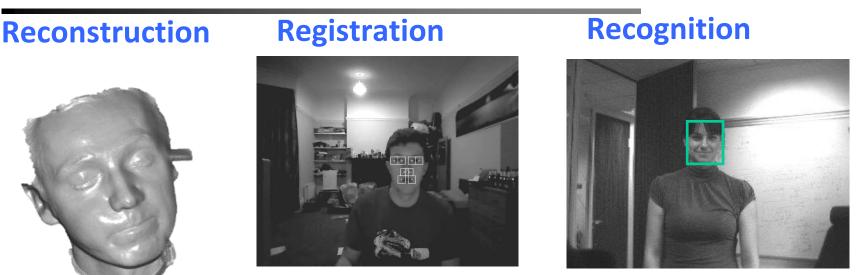
3. Geometry and uncertainty in deep learning



## 2 Computer Vision at Cambridge

## Computer Vision: 3R's





#### **Reconstruction:** Recover 3D shape

#### **Registration:** Compute their position and pose

#### **Recognition:** Identify objects

## Computer Vision: 3R's



## **Recognition Registration Reconstruction**

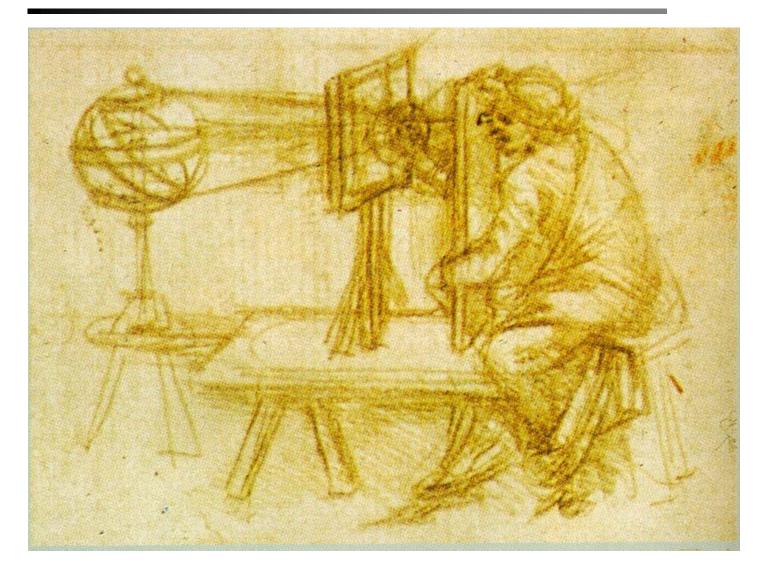
#### **Reconstruction:** Recover 3D shape

#### **Registration:** Compute their position and pose

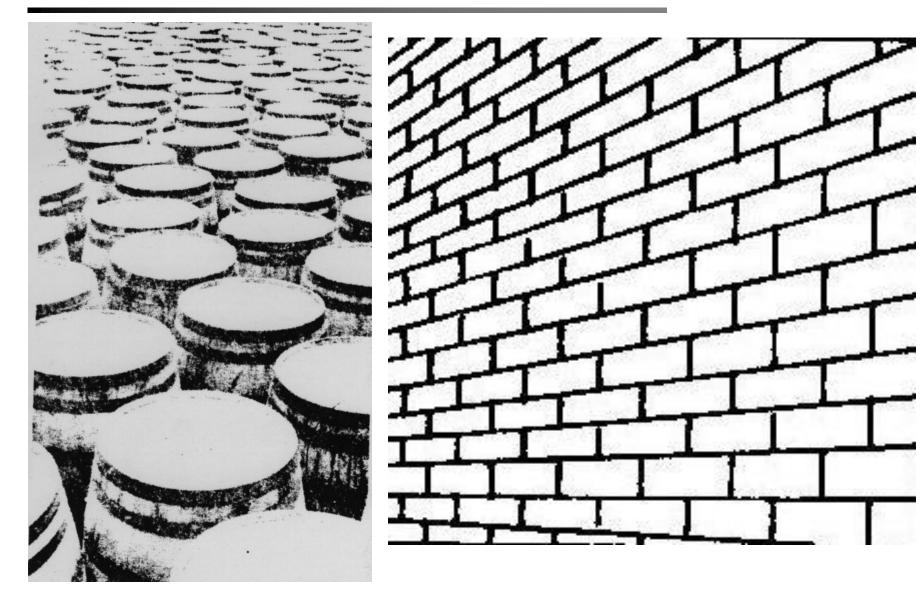
#### **Recognition:** Identify objects

#### 1 Geometry - Perspective



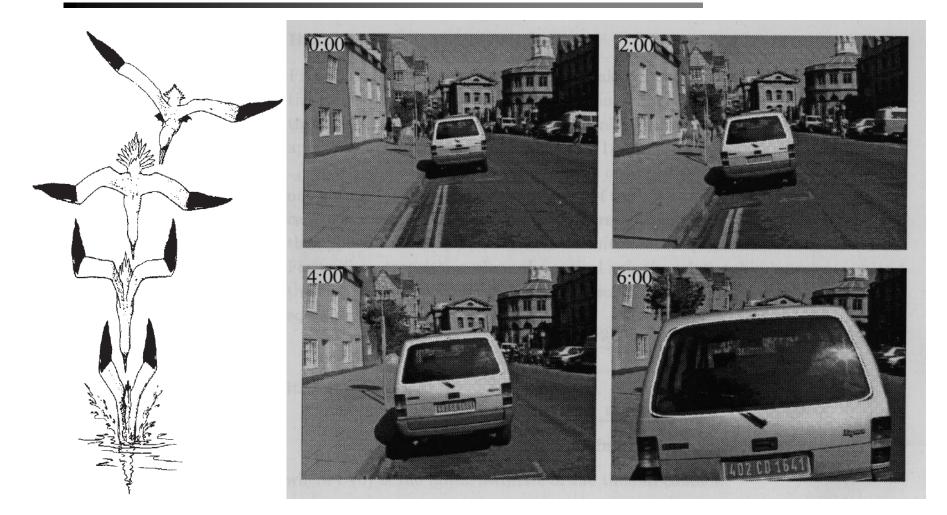


## Geometry - Transformations Suniversity of CAMBRIDGE



#### Time to contact



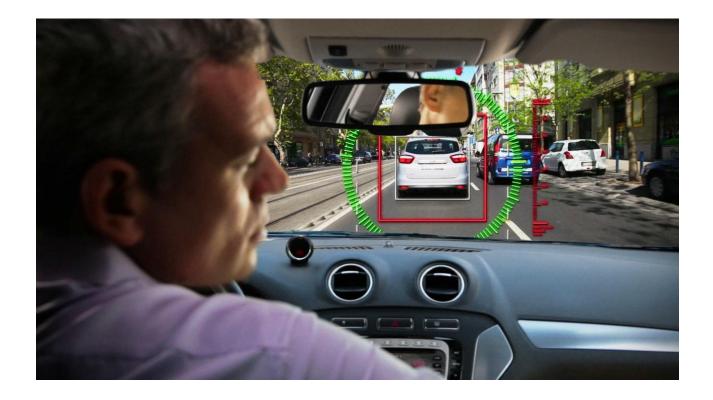


Lee and Reddish 1981

Cipolla and Blake 1992

#### Time to contact







## **Reconstruction?**

# Recovery of 3D shape from images

#### Reconstruction



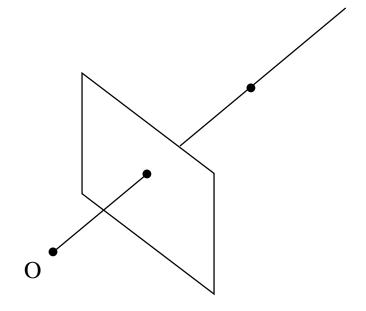






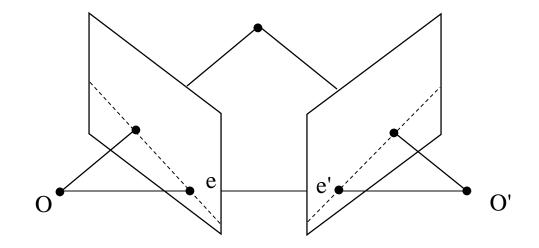
## Ambiquity in a single view





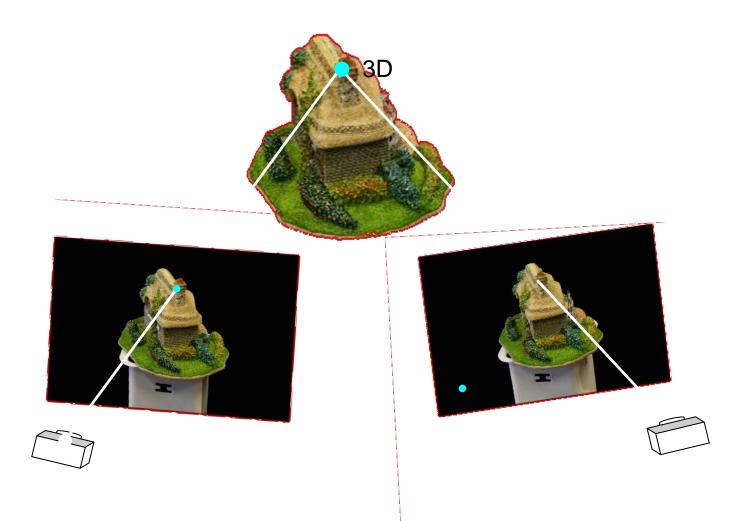
#### Stereo vision





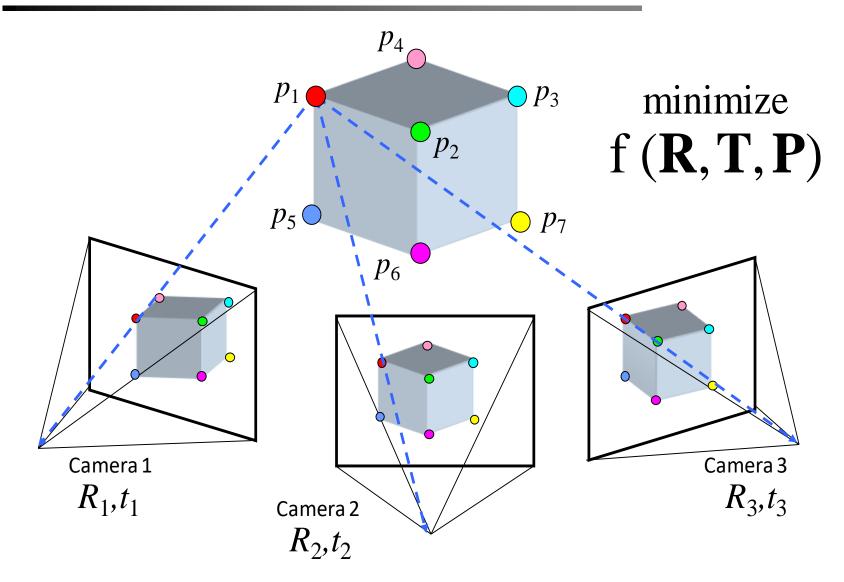
#### Stereo vision





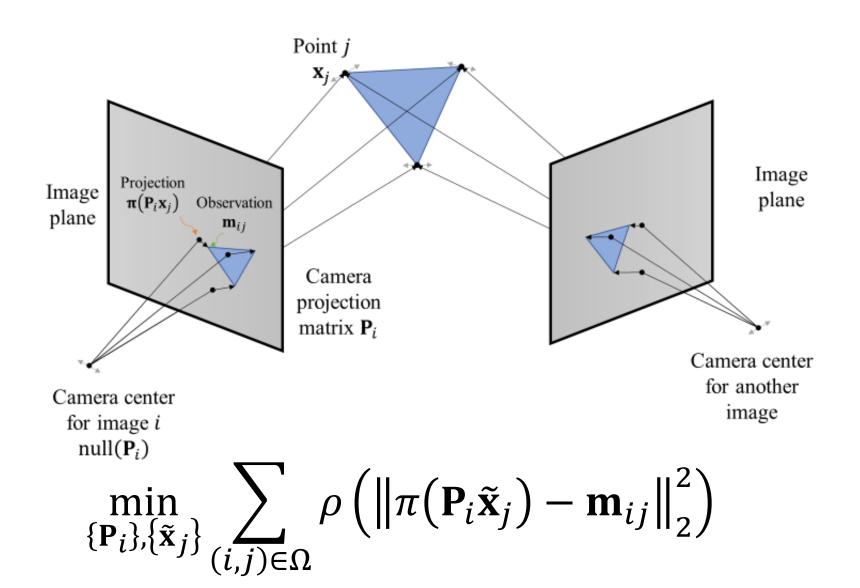
### Multi-view stereo





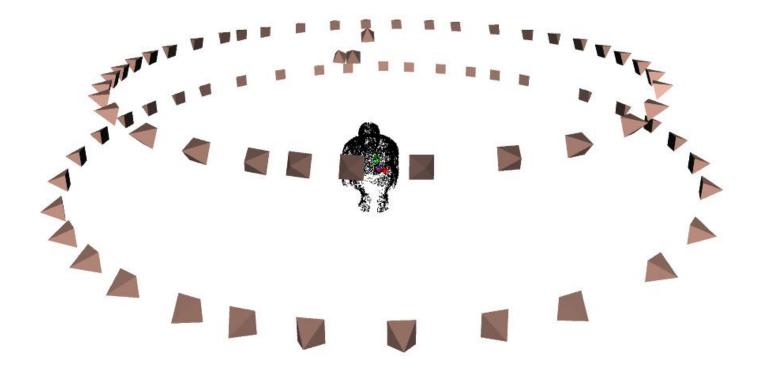
## SFM – bundle adjustment





## Motion estimation result





**Cipolla and Giblin 2000** 

### 3D Models – Gormley



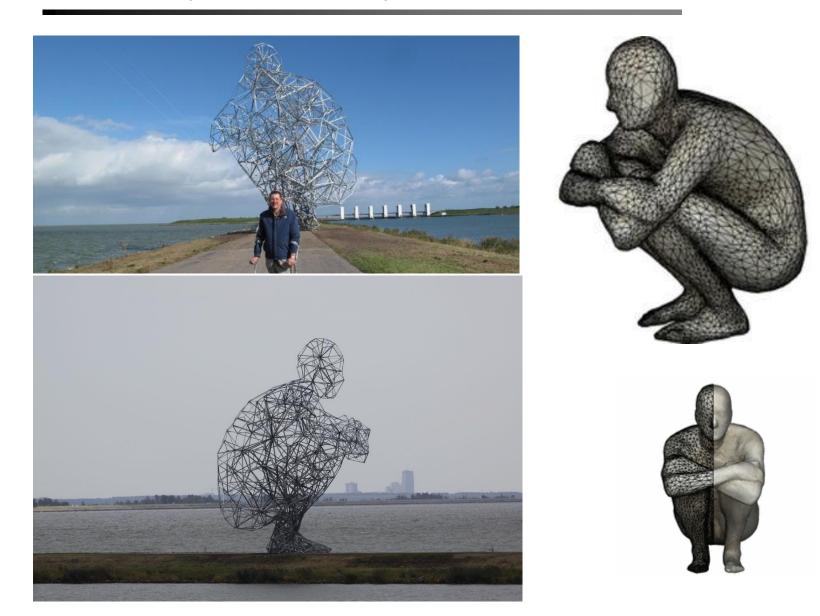


#### **Cipolla and Giblin 2000**

Hernandez and Cipolla 2005

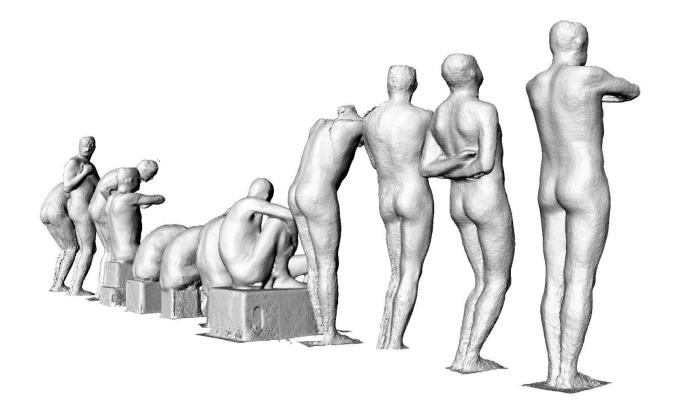
## Antony Gormley





## 3D scans - Antony Gormley







## Multiview photometric stereo

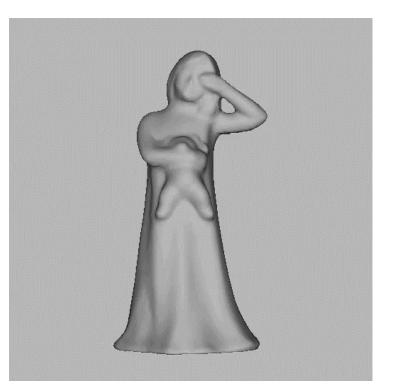
Vogiatzis, Hernandez and Cipolla 2006 and 2008

### Photometric stereo



#### • Assumptions:

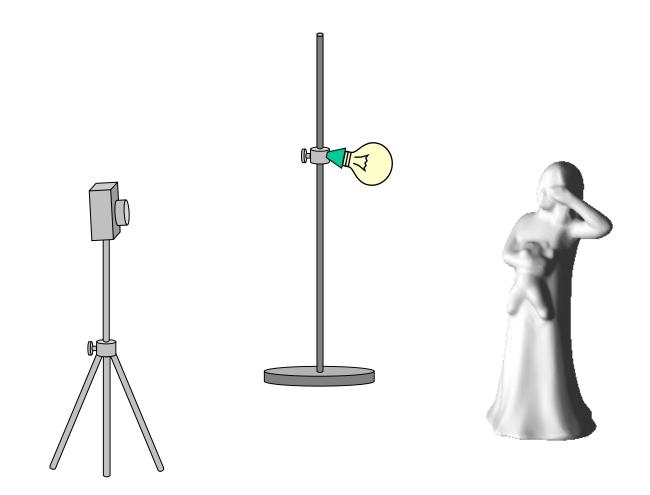
- Single, distant light-source
- No texture, single colour
- Lambertian with few highlights



#### Changing lighting uncovers geometric detail

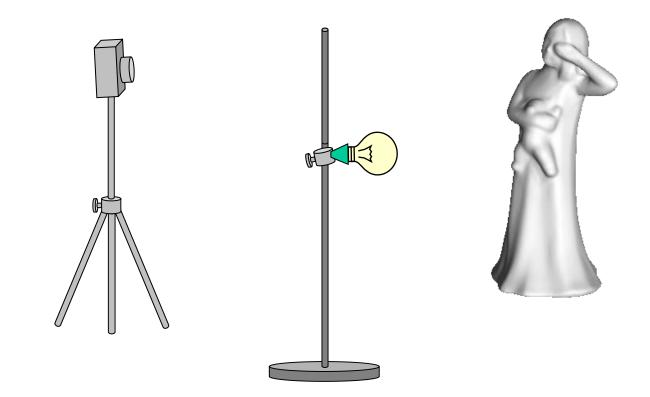
### Classic photometric stereo





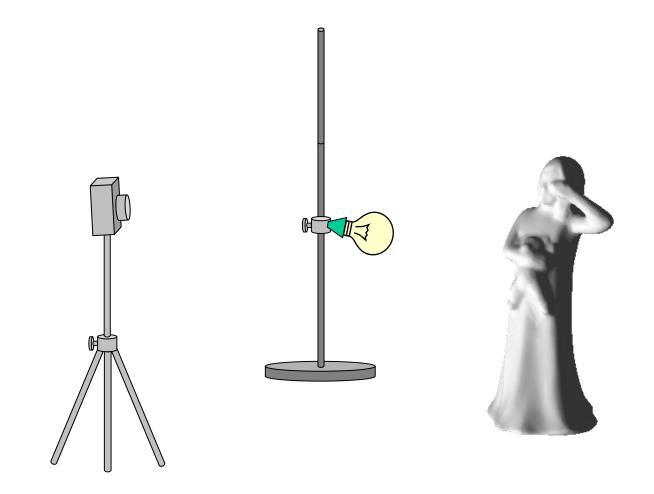
### Classic photometric stereo





### Classic photometric stereo





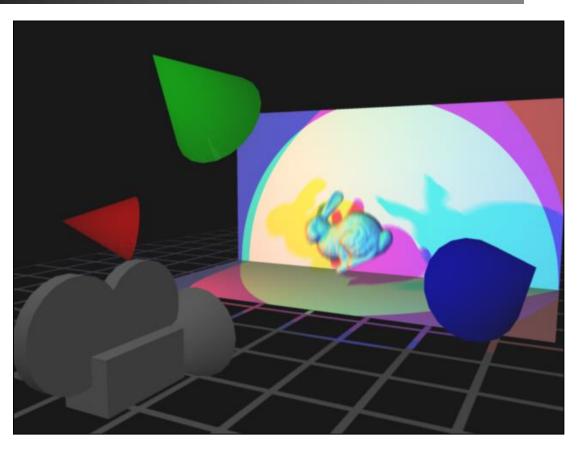


### Deformable objects:

# Real-time photometric stereo using colour lighting



### Textureless deforming objects



a method for reconstructing a textureless *deforming* object in 2.5d

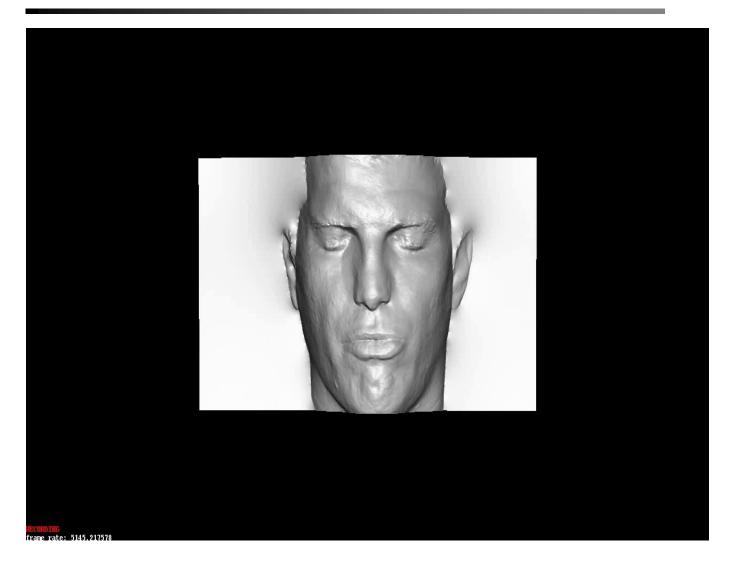
### **Colour Photometric Stereo**





#### Real-time deformable surfaces

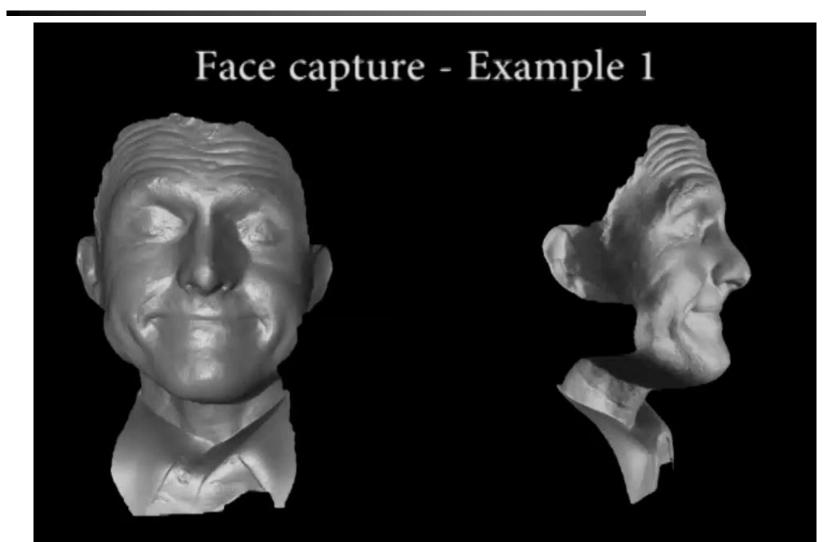




#### Hernandez, Vogiatzis and Cipolla 2008

#### **Sample Reconstructions**





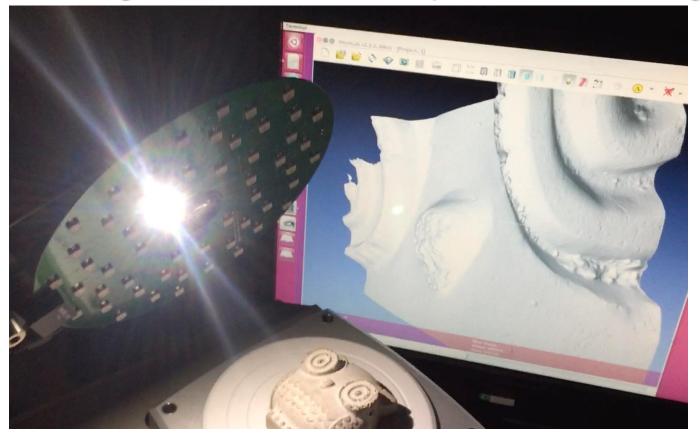
Original viewpoint

Novel viewpoint



### **3D Flash Light**

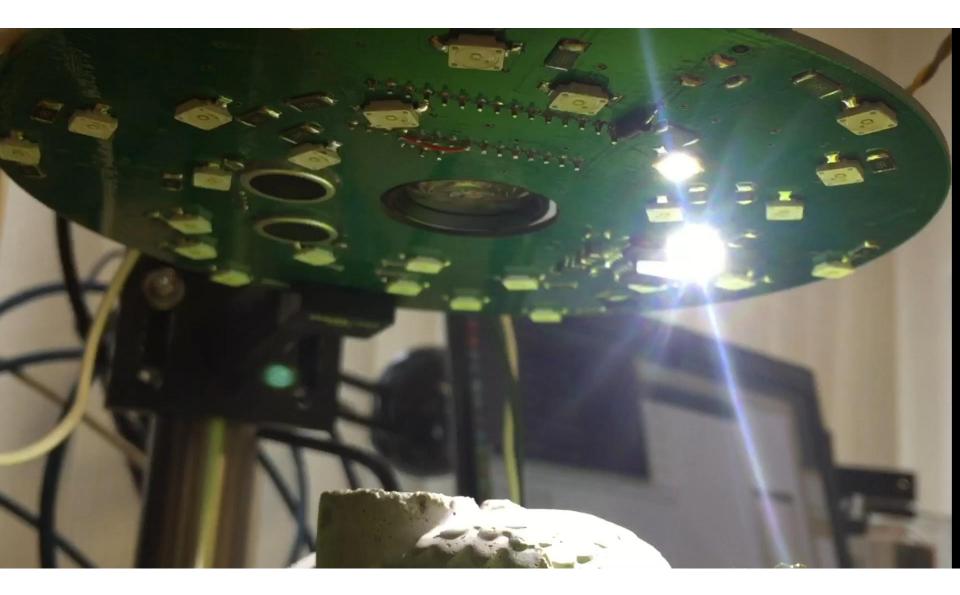
Fast, high resolution and inexpenve 3D scanning technology



#### Mecca, Logothetis and Cipolla 2017



#### 3D flash-light prototype



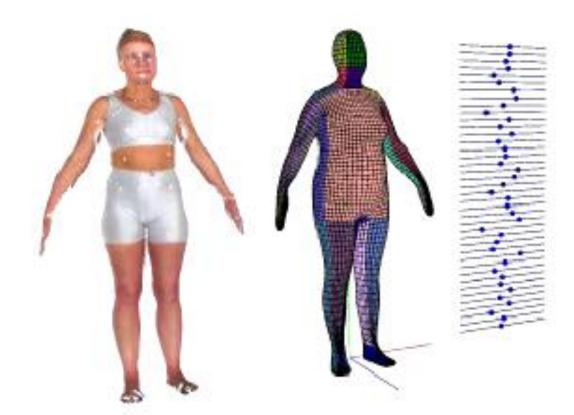
### 2 Dealing with ambiguity





Ames (1946) Room

#### Dealing with incomplete sensory data



Helmholtz (1866)-

"Perception is our best guess as to what is in the world, given our current sensory input and our prior experience."





Probabilistic framework to understanding vision and for building systems:

- 1. Deal with the ambiguity of the visual world
- 2. Are able to fuse information
- 3. Have the ability to learn

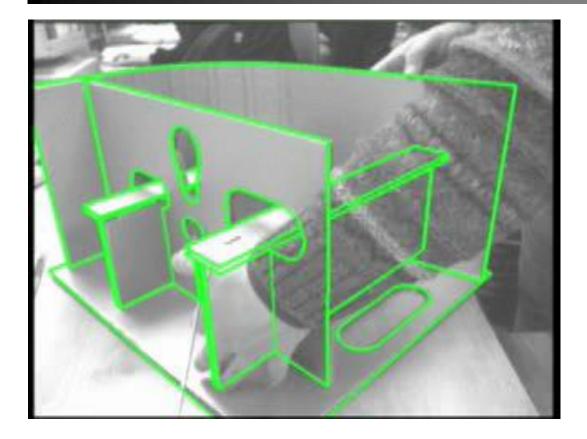


# **Registration?**

# Target detection and pose estimation

### 3D model-based tracking

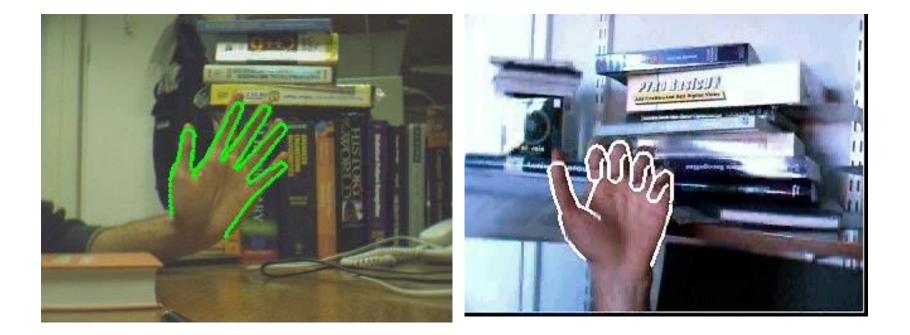




#### **Drummond and Cipolla 2002**

#### **Template-based detection**

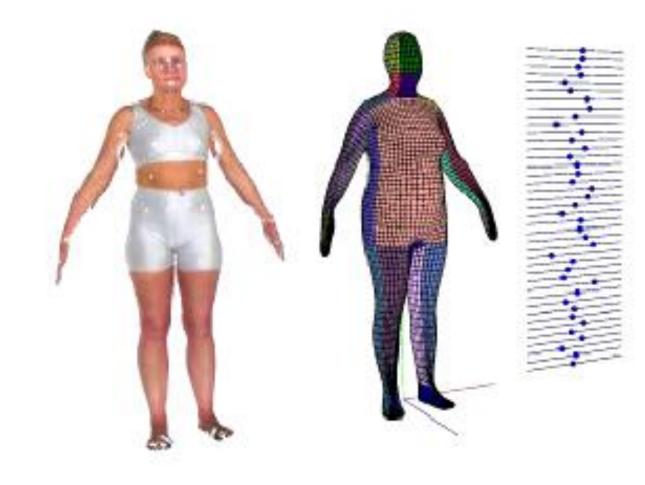




#### Stenger, Thayanathan, Torr and Cipolla 2006

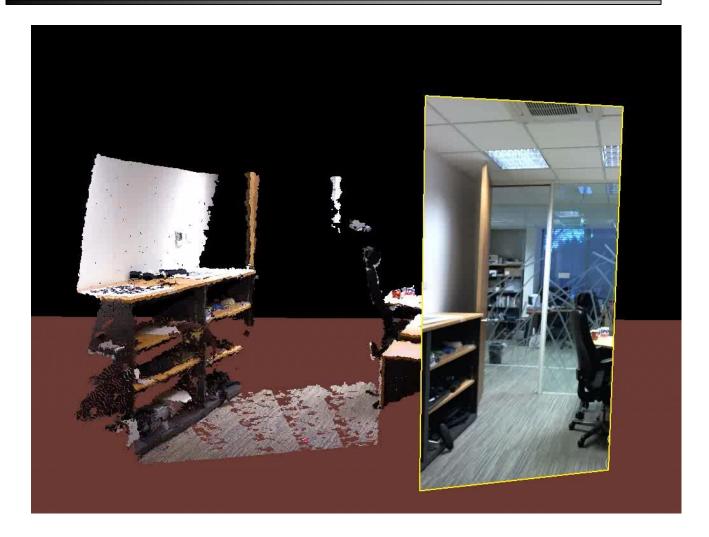
### **Registration – Body shape**





## Single-shot Body Shape





## Single-shot Body Shape







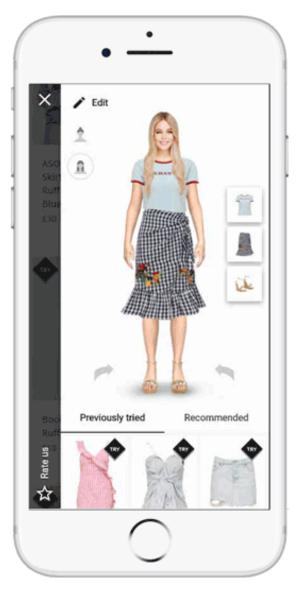
# Single-shot Body Shape





### 3D Registration – magic mirrors







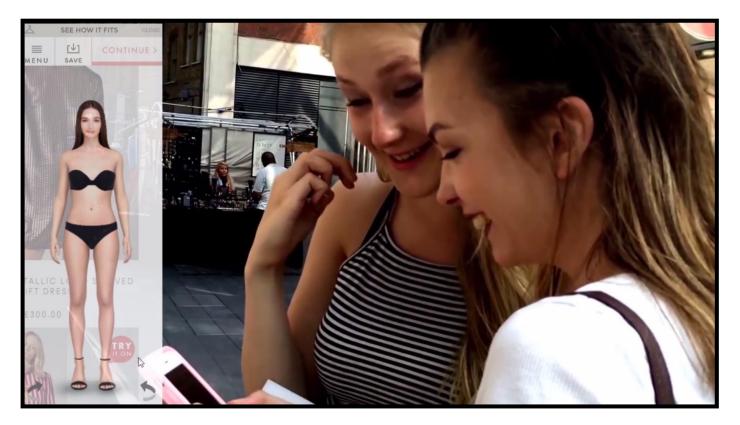




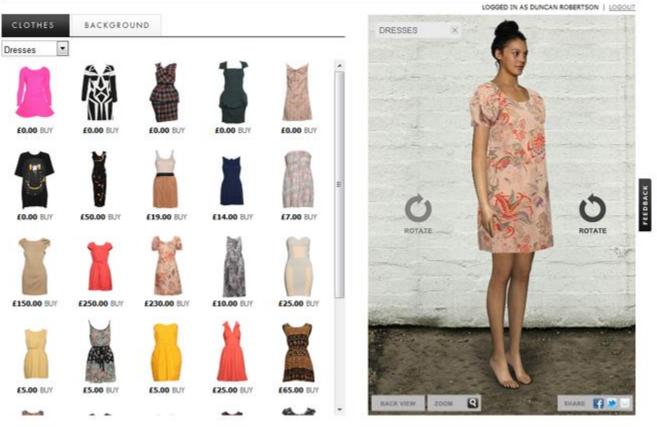
https://trymetail.com



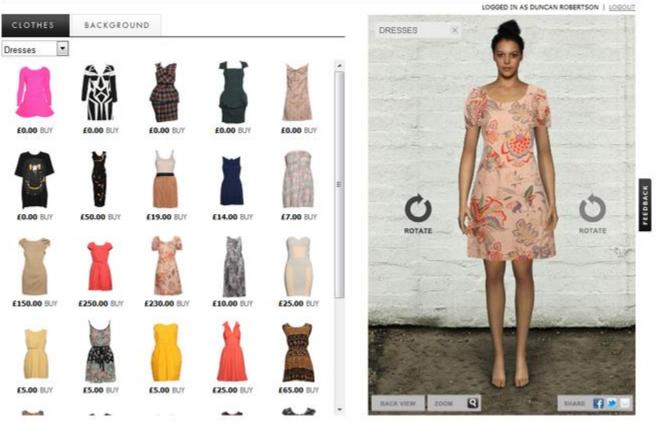
#### Metail Experience: Virtual Try-On



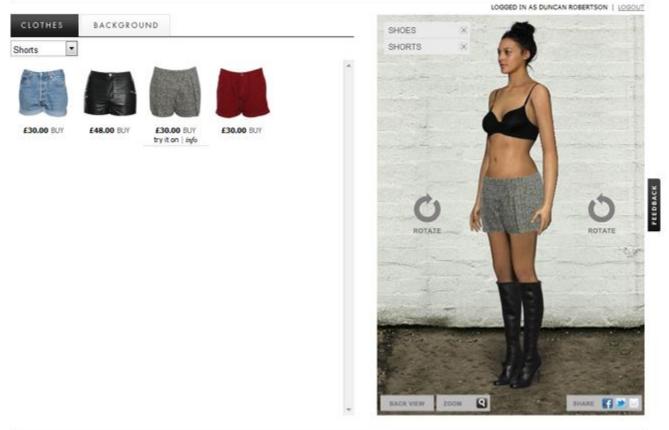




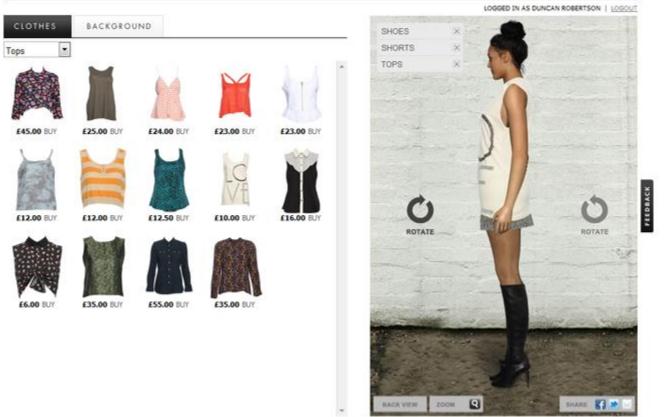




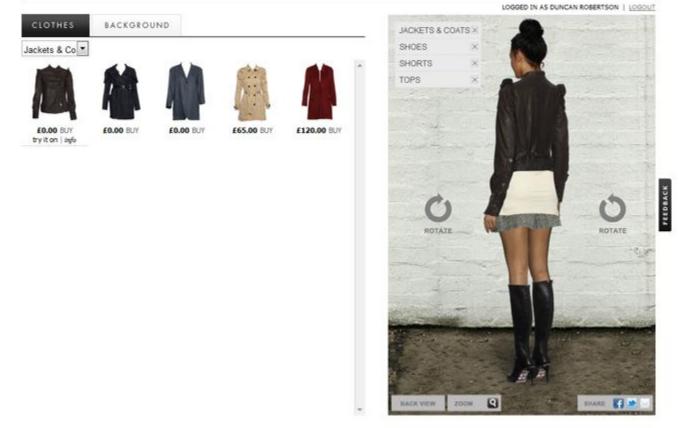












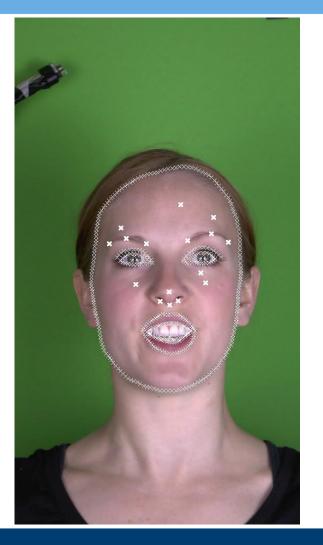


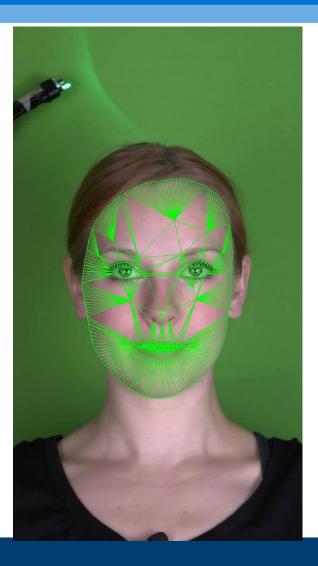
# **Registration:**

# Expressive Visual Text-to-Speech

Anderson, Stenger, Wan and Cipolla 2013

#### **Registration – alignment of training data**



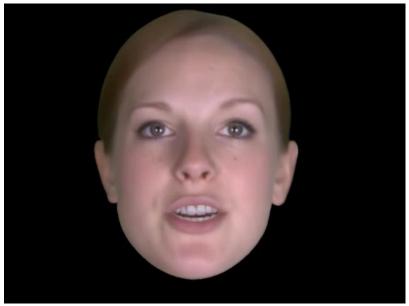




#### What is an expressive talking head?

> User inputs a sentence which they wish to be uttered> User specifies an emotion

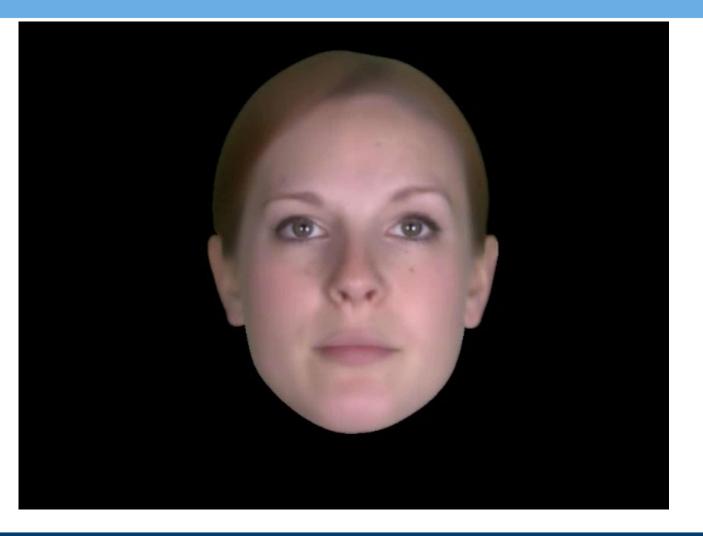
#### Video output is generated







#### **Our current talking head**









#### **Expressive Visual Text to Speech**

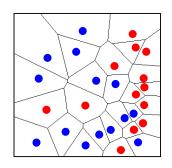


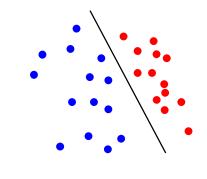


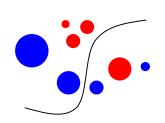


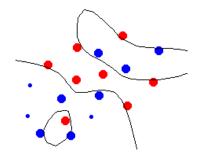
### **3 Machine Learning**



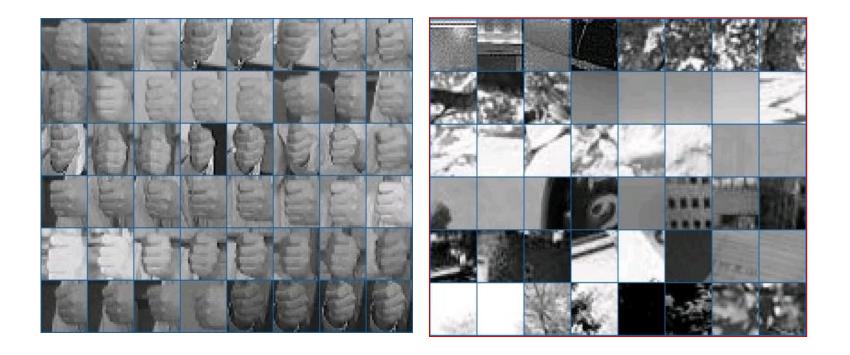








#### Training data – supervised learning



#### Real-time classifiers - Hand detection (2006)

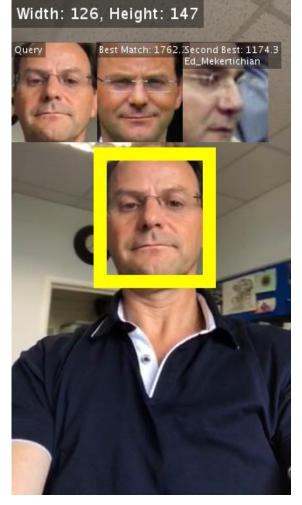


#### Stenger et al. 2006

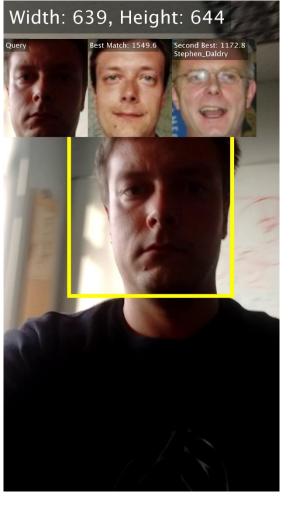


#### Face recognition – similarity learning







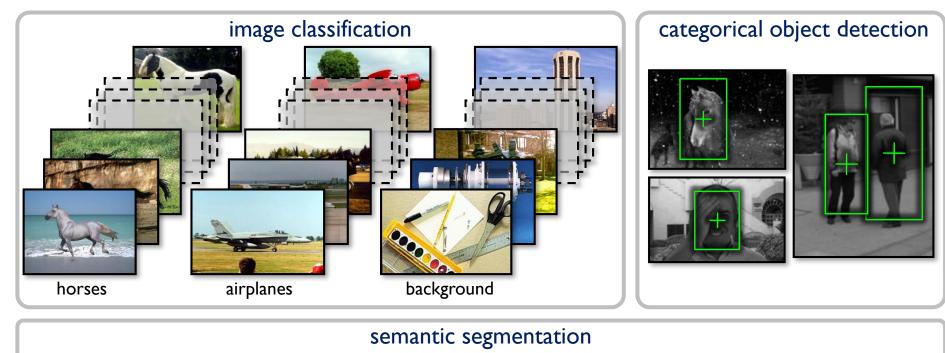




# Recognition?

### Recognition

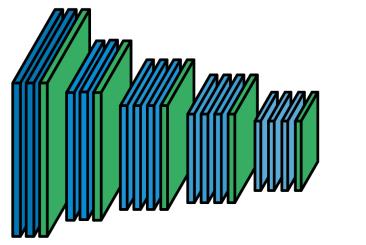








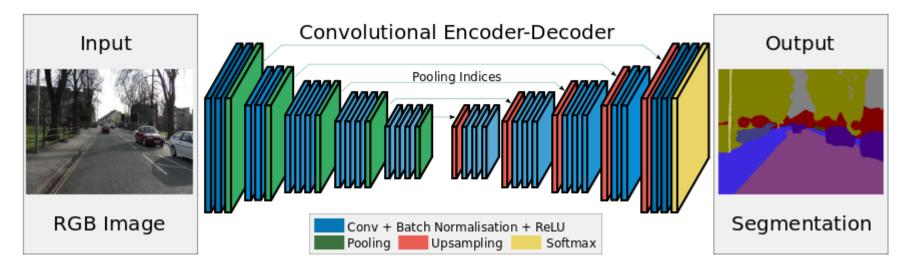




I think this is a "car"

- Input data rich in spatial dimensions
- Output deep features with pooled spatial dimensions

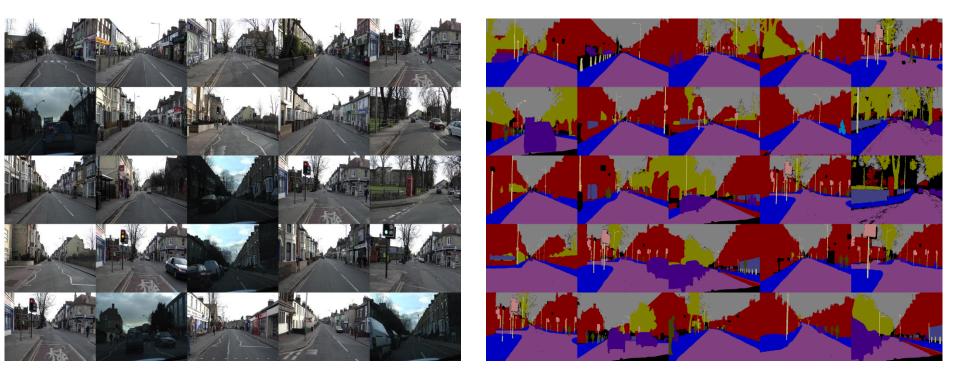




- Real time deep encoder-decoder architecture
- 26 layers, trained end-to-end using stochastic gradient descent
- Inference time is <30ms for a single image on a GPU

#### Badrinarayanan, Kendall and Cipolla 2015 and 2017

#### SegNet – training from labelled data



#### **Brostow, Fauqueur and Cipolla 2009**



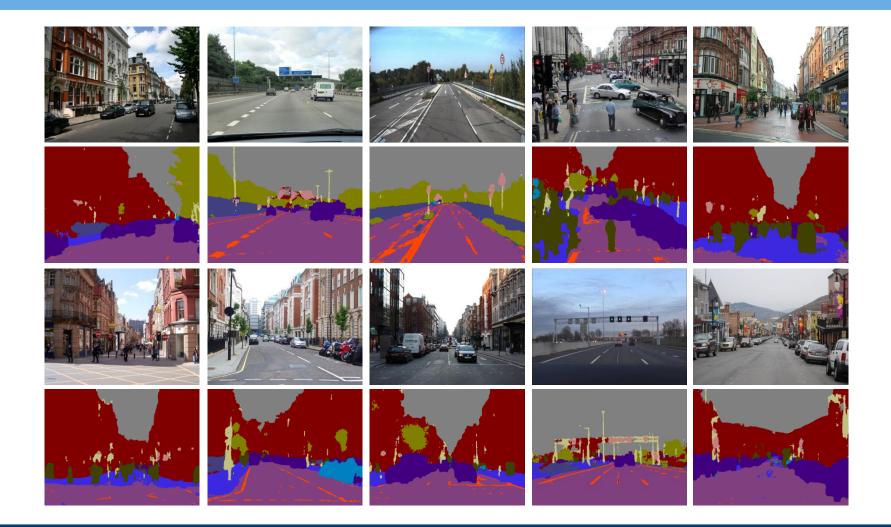
# CamVid Dataset – 21K Images, 700 labelled images, 960x720



#### **Brostow, Fauqueur and Cipolla 2009**

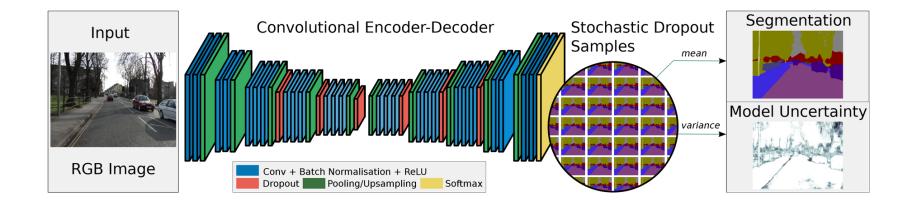


#### SegNet predictions on unseen test images - DEMO





#### **Bayesian SegNet**

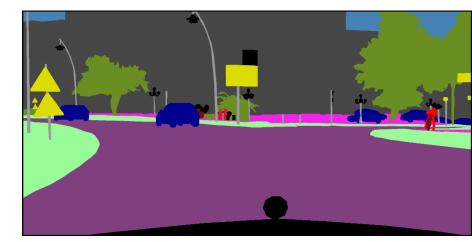


#### Kendall, Badrinarayanan and Cipolla 2017



# CityScapes Dataset – 150K Images, 5K labelled images, 2048x1024







### Label propagation

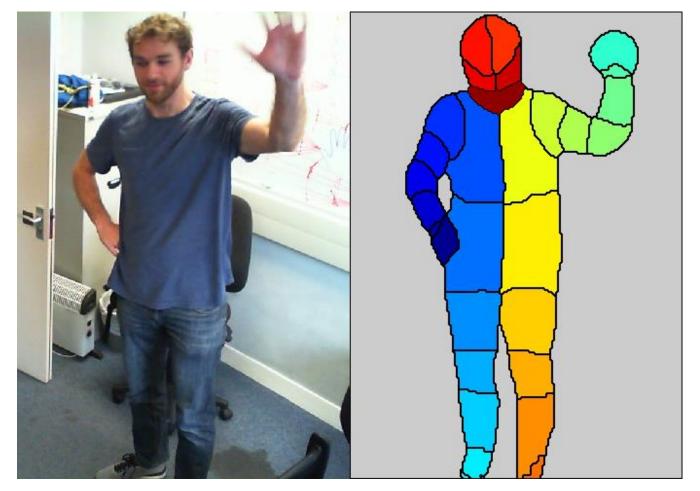




Badrinarayanan, Budvytis and Cipolla 2013

### **Real-time body segmentation**

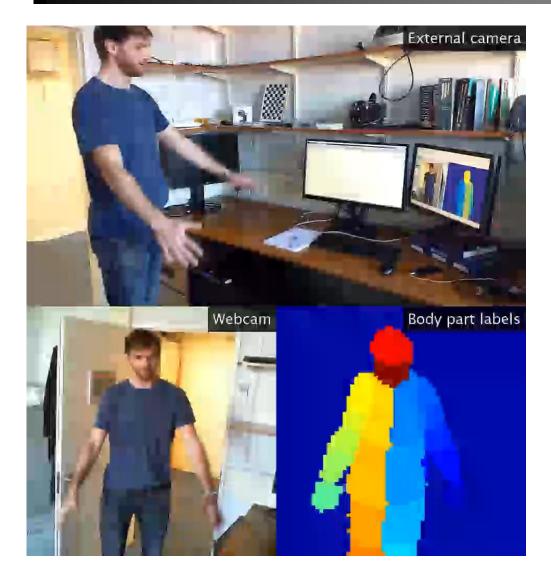




**Charles, Budvytis and Cipolla 2017** Real-time factored convnets (31 body parts at 11fps)



# Real-time body segmentation





# Real-time body segmentation







### **Clothing segmentation**





### **Clothing segmentation**



#### Live demo on iPhone





- CPU Quad-core 2.34 GHz CPU
- GPU PowerVR Series7XT Plus (six-core graphics)
- Memory 3GB (CPU and GPU shared)
- Back facing cameras both 12MP but different lenses
- Accelerometer

Write Apps in Swift or Objective-C (or both)

Various libraries useful for computer vision (BLAS, LAPACK, Vision, ARKit).

Metal framework – GPU-accelerated processing.

CoreML – library for handling computation of deep nets.



### 3 Geometry and Uncertainty in Deep Learning



If we can come up with an effective parameterisation And we can generate large datasets Then, we can learn powerful, real-time perception systems using supervised deep learning

Deep learning architectures are effective for classification problems, even at an image pixel level

Using Monte Carlo Dropout we can estimate the model's uncertainty

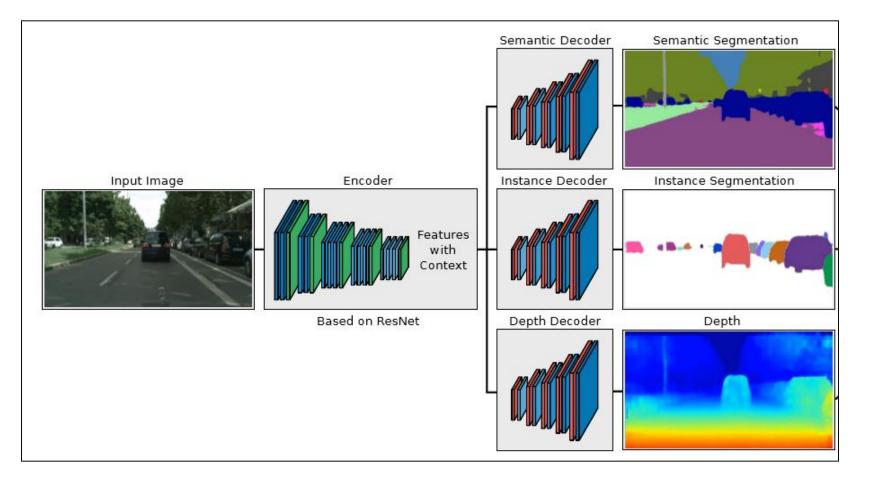


Generalisation against goals, tasks and environments

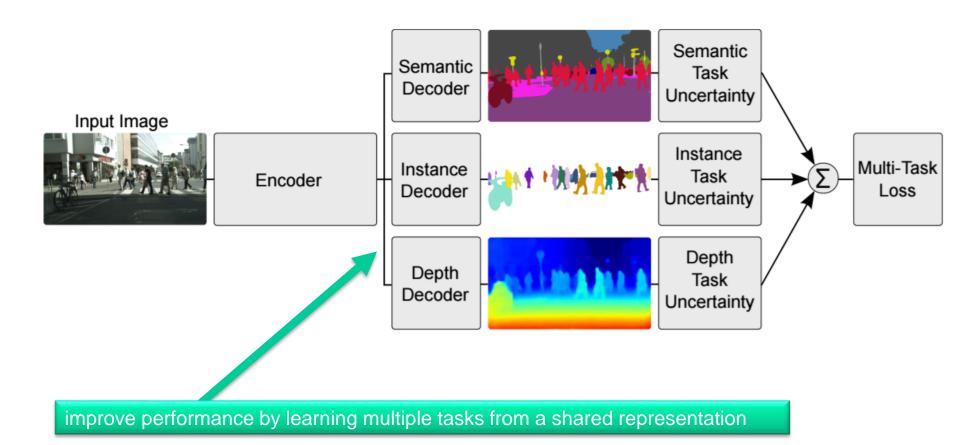
- Multi-task and multi-modal:
  - share representations, avoid being task-specifc
- Data augmentation
  - Transfer learning, domain adaptation and CG
  - Distance metric learning -> one-shot learning
  - use Uncertainty, Geometry and Physics
- Video ability to predict
  - Next big challenge, better learned representations

### Multi-Task Learning



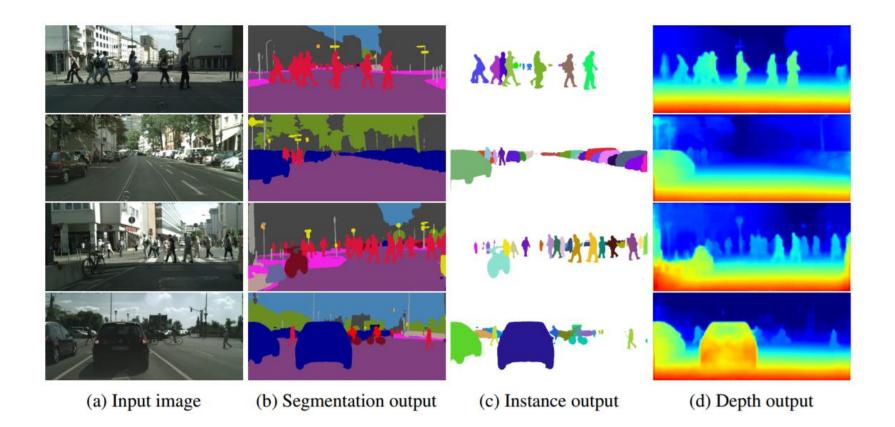


### Multi-Task Loss and Uncertainty CAMBRIDGE



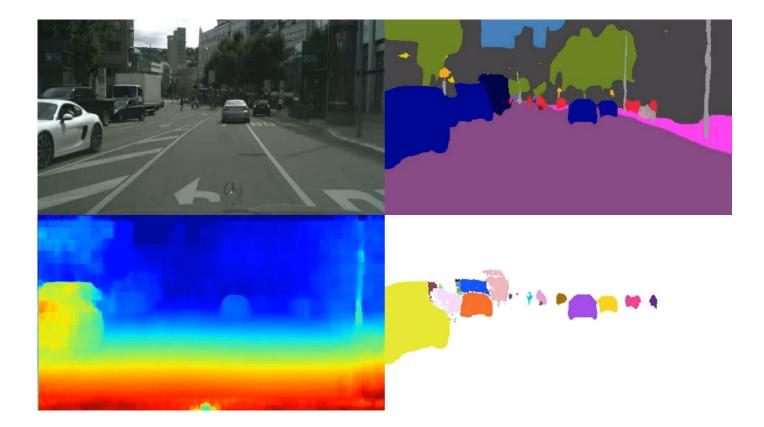
### **Multi-Task Learning**





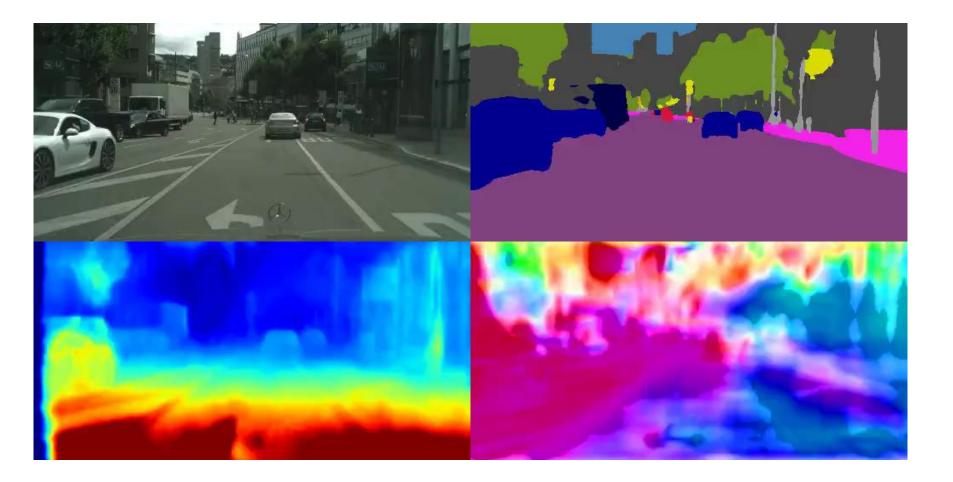
### Multi-Task Learning





## Semantics, geometry and motion Semantics

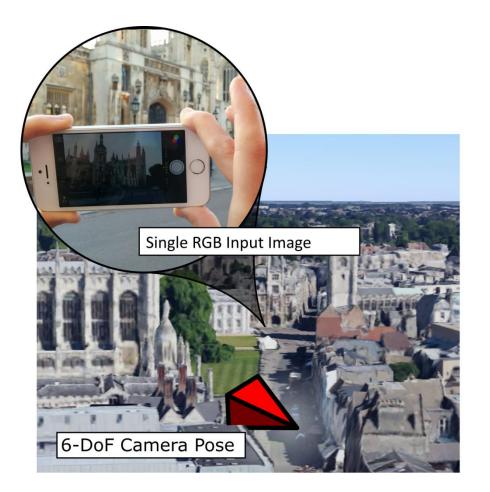




Kendall and Cipolla 2018 VideoSegNet – learning motion and geometry for video semantic segmentation

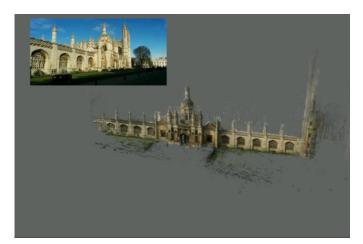


## PoseNet -Realtime 6-DOF Camera Relocalisation



### Where am I?

- camera relocalisation,
- loop closure,
- robot kidnap problem...





Modelling Uncertainty in Deep Learning for Camera Relocalization. Alex Kendall and Roberto Cipolla. | <u>mi.eng.cam.ac.uk/projects/relocalisation</u>/

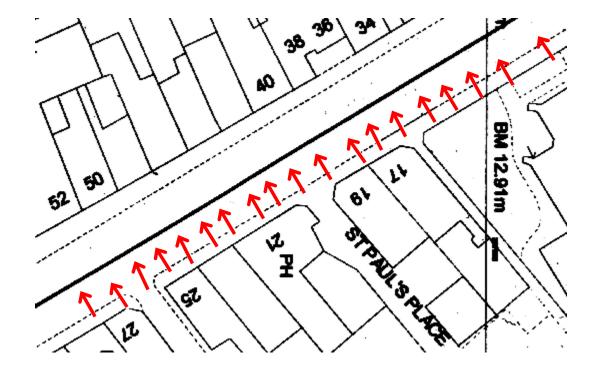
### Image-based localisation





### Image-based localisation







### Image-based localisation



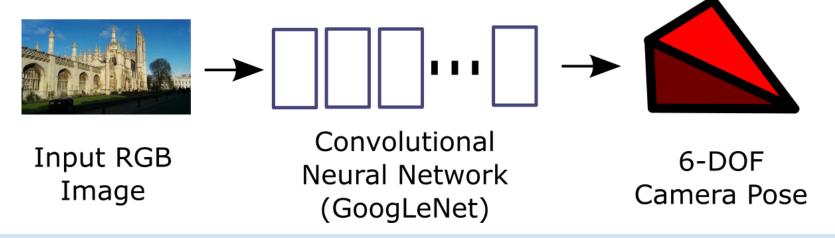


#### **Robertson and Cipolla 2004**



### PoseNet

Learns to regress camera's position and orientation using supervised deep learning.



**PoseNet: A Convolutional Network for Real-Time 6-DOF Camera Relocalization.** Alex Kendall, Matthew Grimes and Roberto Cipolla Proceedings of the International Conference on Computer Vision (ICCV), 2015.

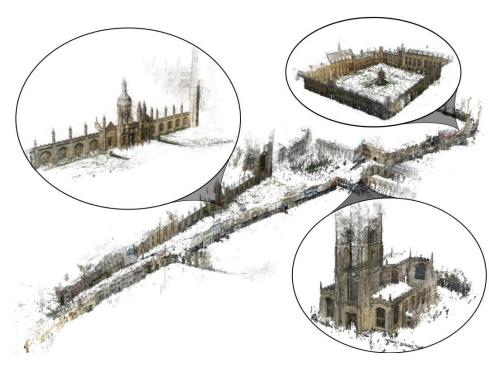


Modelling Uncertainty in Deep Learning for Camera Relocalization. Alex Kendall and Roberto Cipolla. | <u>mi.eng.cam.ac.uk/projects/relocalisation</u>/

### **Cambridge Landmarks Dataset**

- 10,000+ images
- Structure from motion to label images with their global pose
- Train and test images from distinct sequences

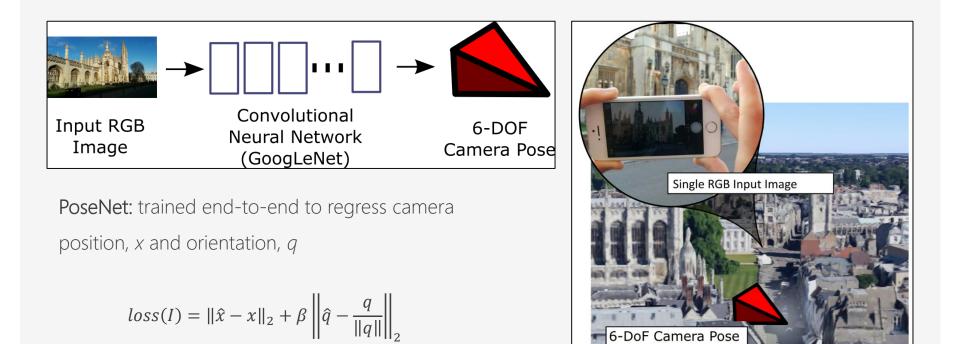






Modelling Uncertainty in Deep Learning for Camera Relocalization. Alex Kendall and Roberto Cipolla. | <u>mi.eng.cam.ac.uk/projects/relocalisation</u>/

#### Naïve deep learning approach to learning camera pose

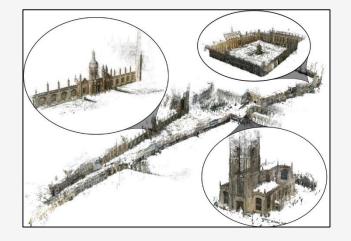


#### Kendall, Grimes and Cipolla 2015

Train with reprojection loss of 3-D geometry with predicted and ground truth camera poses.

$$loss(I) = \frac{1}{|\mathcal{G}'|} \sum_{g_i \in \mathcal{G}'} \|\pi(\mathbf{q}, \mathbf{x}, \mathbf{g_i}) - \pi(\mathbf{\hat{q}}, \mathbf{\hat{x}}, \mathbf{g_i})\|_{\gamma}$$

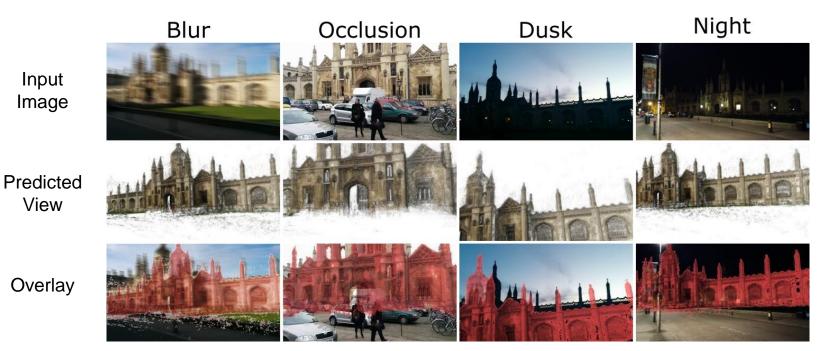
Where  $\pi$  is the projection function of 3-D point  $g_i$ 



#### Kendall and Cipolla 2017

### Deep Learning Robustness

The system is tolerant to conditions very different from those in its training set

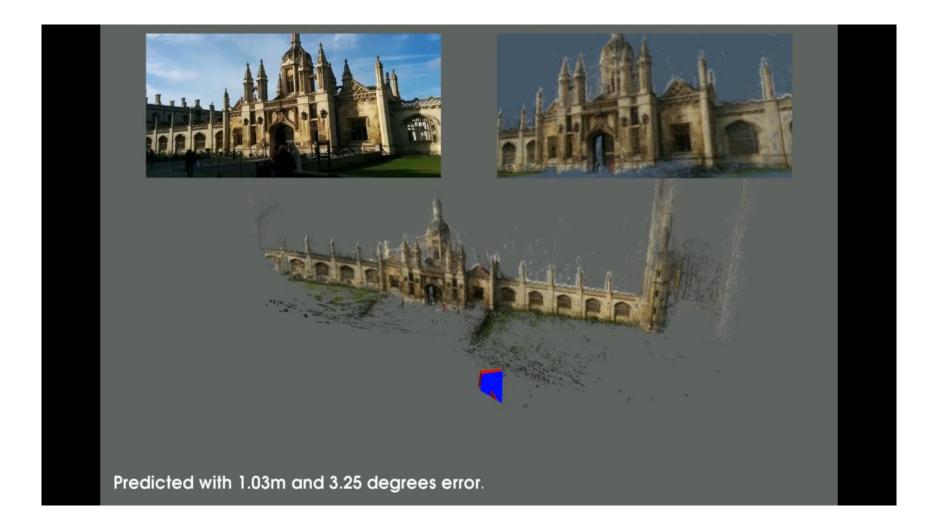




Modelling Uncertainty in Deep Learning for Camera Relocalization. Alex Kendall and Roberto Cipolla. | <u>mi.eng.cam.ac.uk/projects/relocalisation</u>/

### Visual relocalisation





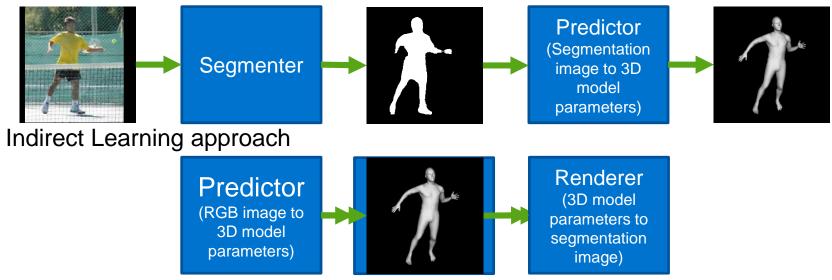


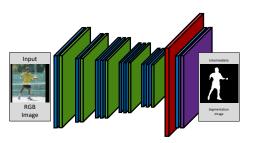
# Structured Learning of 3D Shape

Tan, Budvytis and Cipolla 2017

#### **Indirect Learning**

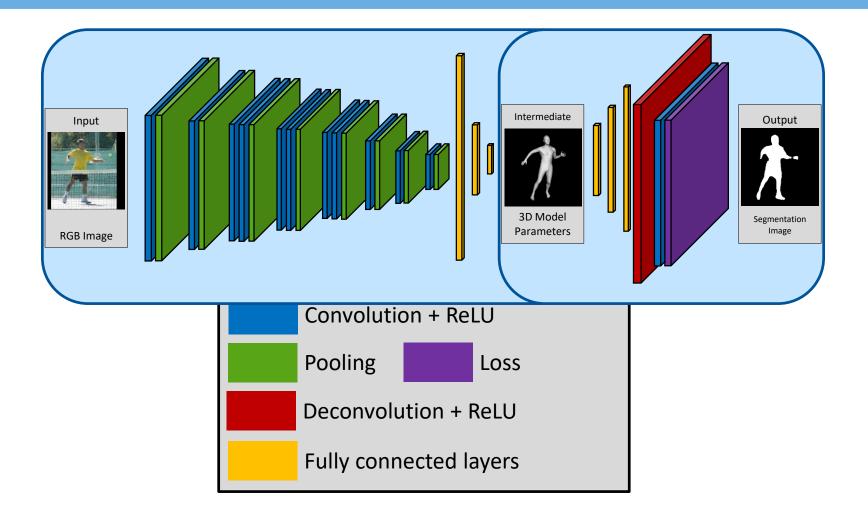
Stacked Networks approach



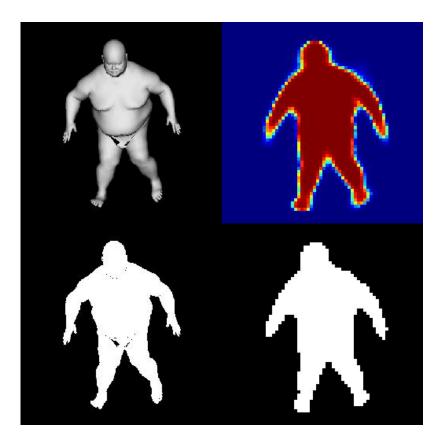




•

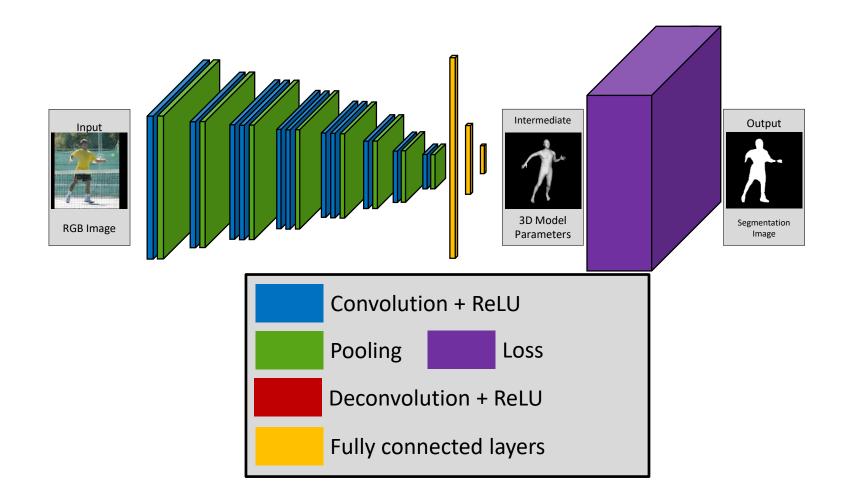




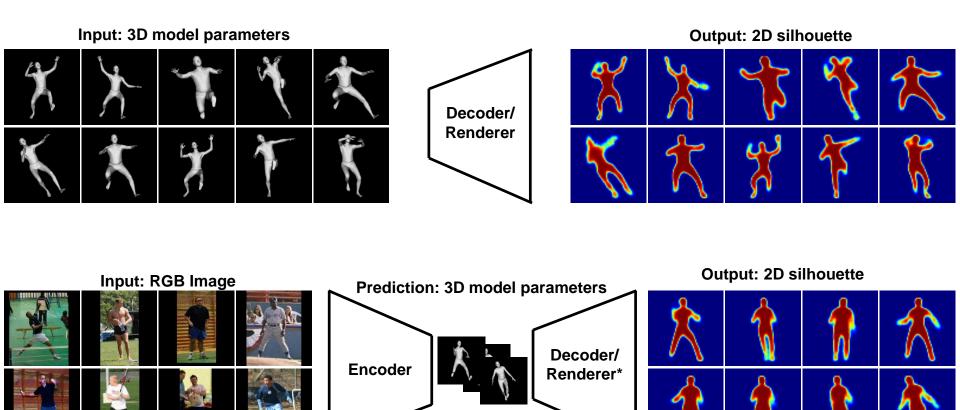


#### Tan, Budvytis and Cipolla 2017









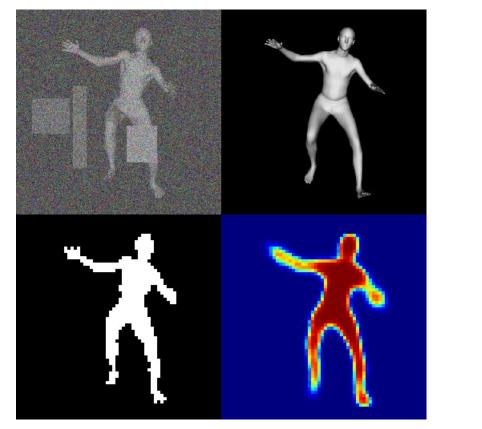
\* Fixed when training full network

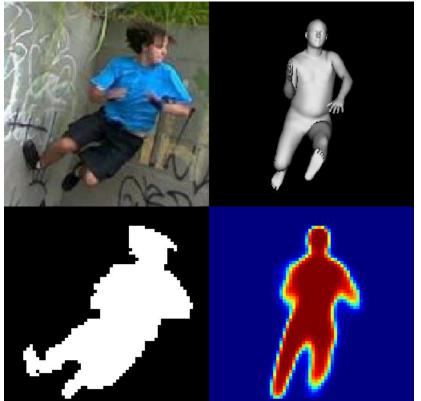




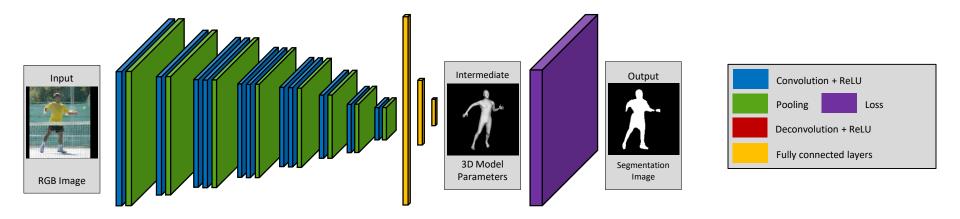
#### Tan, Budvytis and Cipolla 2017





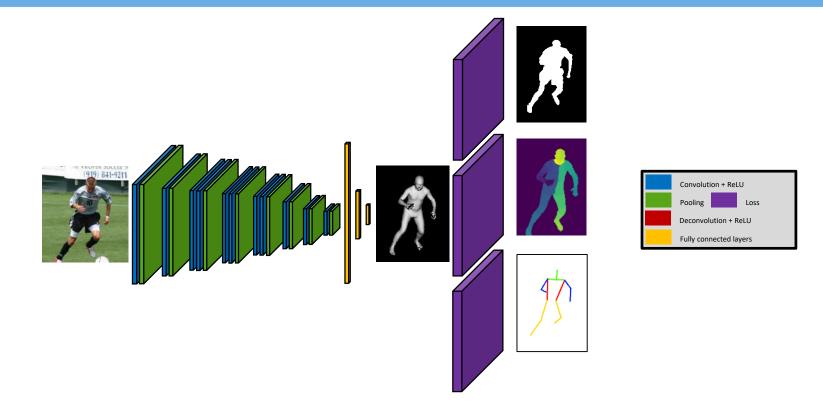






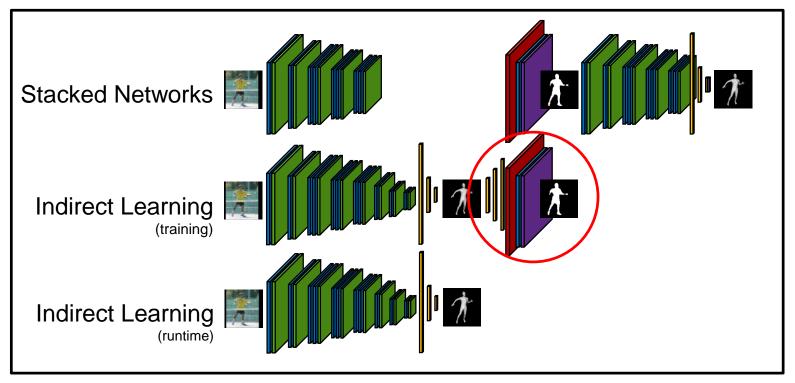
- Easily specified cost function
- Necessarily valid intermediate representations
- No loss of relevant information





• Ability to use multiple data modalities





• Lightweight





- 1. Background: Why and How?
- 2. 3R's of Computer Vision:
  - Reconstruction
  - Registration
  - Recognition

3. Geometry and uncertainty in deep learning



- 1. Perception and Action
  - vision provides representation
  - embodied systems
- 2. Learning from observation and interaction
  - observe consequences of actions through time and multiple modalities
    - imitation, inverse RL and model-based RL
- 3. Explainable AI
  - interpretability, transparency and limitations



# Publications:

http://mi.eng.cam.ac.uk/~cipolla/publications\_selected.htm

# Research demos and code:

http://mi.eng.cam.ac.uk/projects/segnet/ http://mi.eng.cam.ac.uk/projects/relocalisation/

# **Research Videos:**

https://www.youtube.com/user/ComputerVisionVideos



#### Alex Kendall, Ignas Budvytis and James Charles

Carlos Hernandez, Bjorn Stenger and George Vogiatzis

Rob Anderson, Vijay Badrinarayanan, Yu Chen, Matt Johnson, Duncan Robertson, Jamie Shotton and Simon Taylor





# Metail: Virtual Fitting Room (2008)

http://www.metail.com

# Zappar: Augmented Reality (2011)

http://www.zappar.com

# Dogtooth Technologies: Fruit-picking robotics (2015)

http://www.dogtoothtech.com

# Wayve: Autonomous vehicles (2017)

http://wayve.ai

Cambridge Heartwear: Wearable ECG with AI (2016)

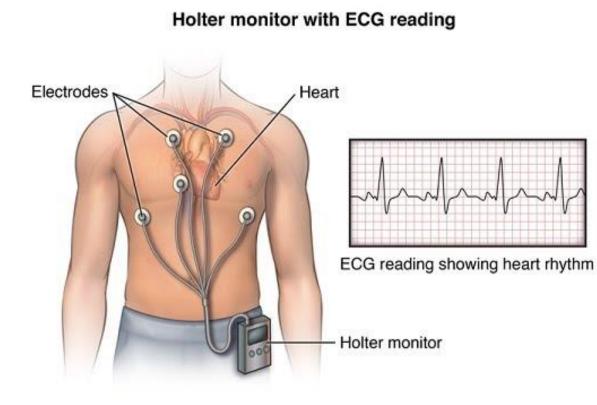




# Cambridge Computer Vision group spin-outs

# Unsupervised learning of ECG





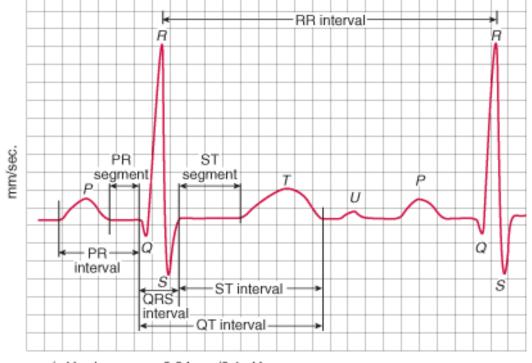
# Unsupervised learning of ECG



[2]

#### Model

- Trained to reconstruct • individual heart beats
- Compresses heart beat signal ٠ from 1024 samples to 10 latent variables



#### .

Dataset

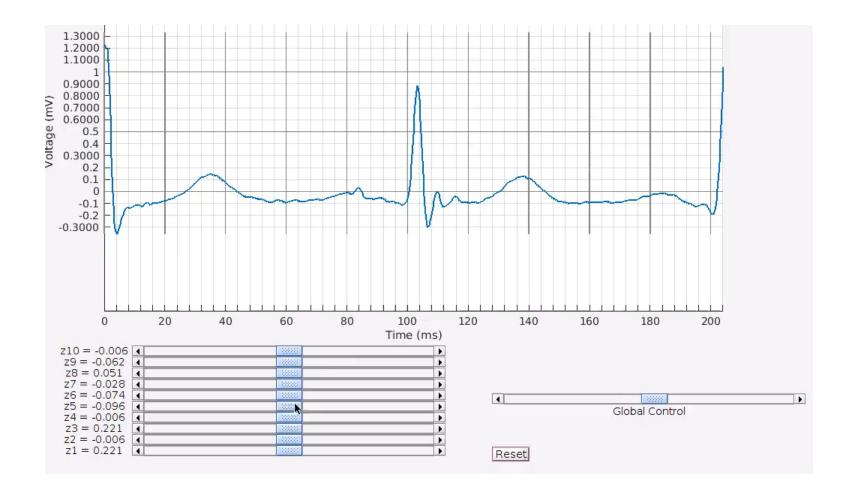
540, 000 individual heart beats extracted from 18 long term (~24hr) ECG recordings of normal rhythm [1]

1 square = 0.04 sec/0.1mV mm/mV

[1] Goldberger A. L., et. al, PhysioBank, PhysioToolkit and PysioNet, Coponents of a New Research Resource for Coplex Physiologic Signals, Circulation, 2000 [2] Mohammed M., et. al, Compression of ECG Signals Based on DWT and Exploiting the Correlation between ECG Signal Samples, Int'I J. of Communications, Network and System Sciences, 2014.

# Unsupervised learning (VAE)







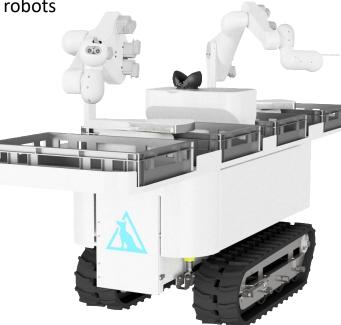




Addressing labour shortage using mass-produced, intelligent robots



One of four v2.0 fruit picking robots



v3.0 design – 24 robots in manufacture 2018

() **Zappar** 

## Zapworks – AR content authoring & publishing platform



# BBC/ Open University - Heart





#### **Create your 3D avatar from basic body** measurements

- Visualize how the clothes fit on you
- **Build layered outfits**
- Provide size & fit advice

#### Backed by computer vision and machine learning technology

- Statistical 3D body shape modelling ٠
- Automated and scalable garment digitisation based ٠ on deep learning
- Garment physics simulation ٠
- High quality photorealistic product image synthesis •
- Intelligent fit advice and recommendation •

#### https://trymetail.com



#### Virtual Try-On





**Composite Photography** 

