



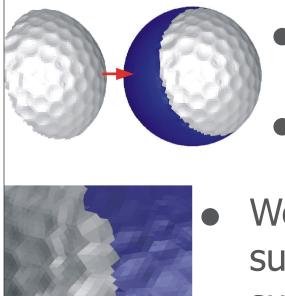
## Overview

At IPAB, we study a variety of problems in perception, action and behaviour from a formal modelling perspective. We use methods such as Bayesian inference, hidden Markov models and statistical learning to model and investigate a broad variety of topics from insect sensing and behaviour to motor control in robotic and biological systems, and computer vision.

## **3D Computer Vision**

3D computer vision involves reasoning about the nature of the 3D world based on scanned 3D data. We currently have 3 systems for acquiring this data: a very high precision laser scanner, a high resolution stereo system and a state-of-the-art video-rate stereo rig.

### **Plausible 3D Surface Completion**



Uni-directional 3D capture devices can only recover the visible portion of an object's surface But we wish to capture the entire object including backfacing and occluded surfaces

• We have been investigating methods for plausible 3D surface completion using geometric completion and surface relief propagation

#### **Video-rate 3D Analysis**

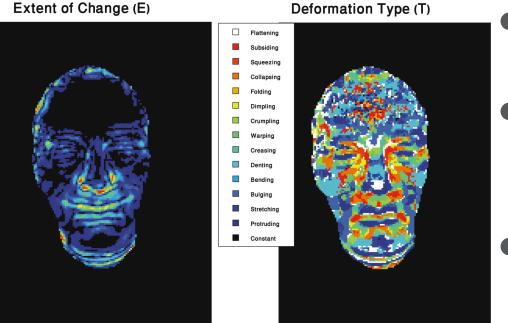




Using surface modelling and statistical techniques, we are attempting to answer questions such as: Can we compactly encode the dynamics of an

- individual's face?
- How unique is this, how does it vary with injury/surgery and can we use it to recover the underlying muscle structure?

#### **Capturing Human Nuance**

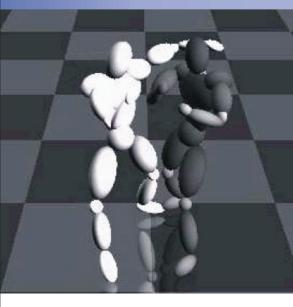


Temporal Characteristics of a smile

Can human nuance be captured by analysing the local curvature of a face?

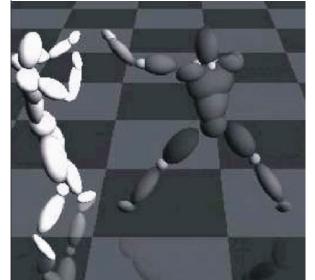
- We are currently extending classification schemes for static surface curvature into the temporal domain
- We seek to identify regions that exhibit variations in shape change (such as the face undergoing expressions), and to characterise the nature of the deformation

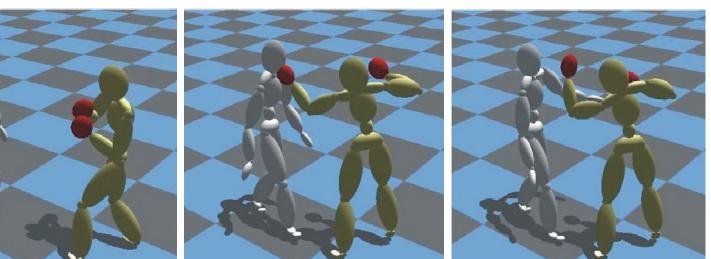
## **Animating Human Avatars**



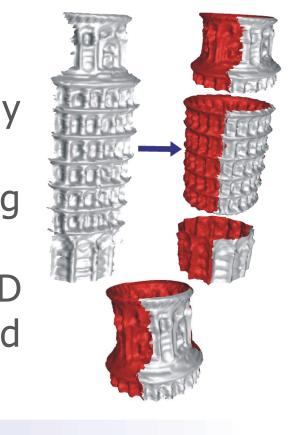
We have developed methods for simulating dense interactions of human avatars, such as pushing, fighting and wrestling. The aim is to generate realistic behaviour in novel contexts based on a small library of motion capture data.

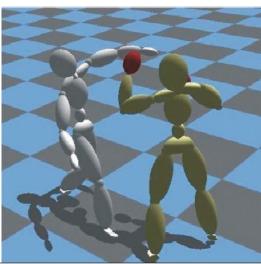
Left: Simulating a fight scene by expanding the game tree Below: Simulating an avatar being hit while walking





# **Institute of Perception, Action and Behaviour**





## **Statistical Learning for Motor Control**

The SLMC group develops statistical machine learning methods, particularly in the context of control of high-dimensional movement systems such as humanoid robots.

### **Sensorimotor Learning in High Dimensional Sys.**

Current research projects include: Robust algorithms for real-time internal model learning in high dimensional movement system Learning cost functions for generating realistic

- human-like movement
- Biological motor learning: Identifying mechanisms and capabilities of learning and representation in the cerebellum.

Left: Learning cost functions for the Honda ASIMO robot (simulated kinematic model is shown here) Right: DLR LW-III dextrous arm with hand

**Learning under Varying Contexts** 



Different situations or contexts require different control strategies: • How can we learn control in a new situation without unlearning what we already know?

- what the context is?
- For example: Is the bottle full or empty?

Left: 7 DOF SARCOS anthropomorphic arm holding a tool

## **Insect Sensing and Behaviour**

The Biorobotics group is interested in studying the behaviour of invertebrates, motivated by the relative simplicity of their nervous systems.

### **Computational and Robotic Modelling**

Models, either physical or simulated, are a valuable tool in investigating theories of behaviour. These models can vary in their level of description from single spiking neurons to complete systems, and are being applied to:

- Phonotaxis for mate localisation in crickets (right)
- Neural network models of insect learning in a classical conditioning paradigm



Multimodal integration: vision and audition in crickets and vision and olfaction in fruit flies (left)

Left: automatically tracked trajectory of flying fly Right: robot with two 'ears', for investigating cricket phonotaxis

## **Biologically inspired robotics**

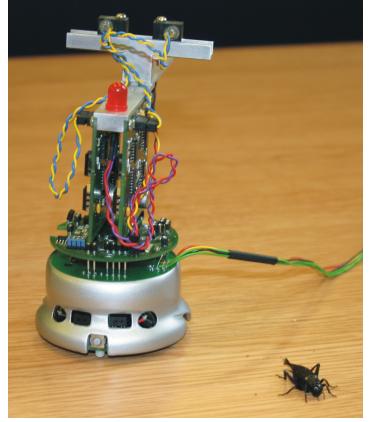
Insights gained from insects can be used to inspire novel algorithms or hardware for robots:

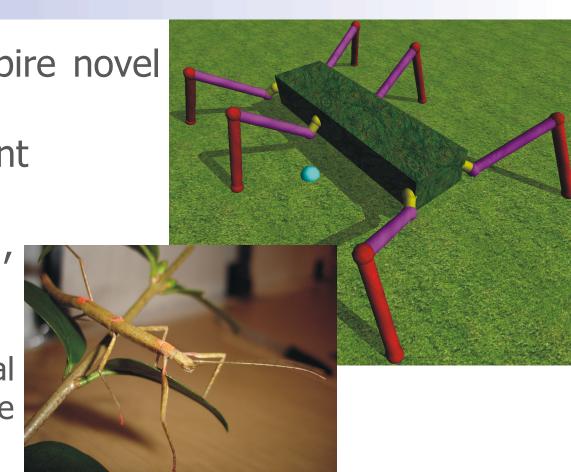
- Landmark-based homing, as seen in the desert ant
- Wind sensing in crickets
- Six-legged locomotion on uneven terrain, inspired by the stick insect (right)

Right: Actual stick insect with coloured markers for visual tracking. Far right: simulated hexapod robot based on the stick insect



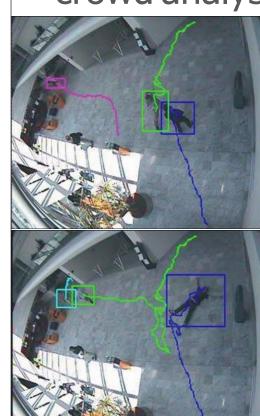
• How can force sensors on the fingertips help to determine





## **Visual Surveillance**

Given the increased demands on security and surveillance system capabilities, and the vast quantity of associated video data, it is essential to automate the various processes involved; from integrating many views of a scene into a single 'superimage', to informing supervisors of interesting events, crowd analysis and behaviour classification.



- interaction

Left: Analysis of an assault captured on a surveillance camera

#### **Detecting Abnormal Crowd Behaviour**

- Flow based analysis of the behaviour of many individuals using crowd-flux statistics and congestion analysis
- We achieve this using explicit models of normal flow using optical flow with Hidden Markov Models used to filter normal crowds

## **Human Multimodal Perception**

We perform psychophysics experiments to find out what people perceive when they receive conflicting information from multiple sources, and we compare the results with a formal Bayesian model of multimodal perception and cue integration.

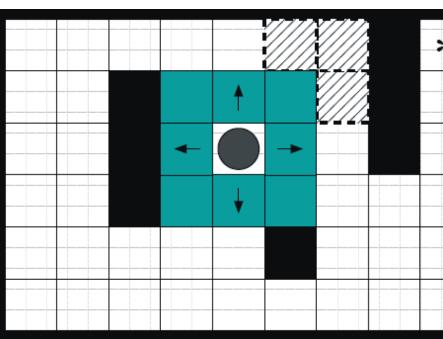
- Do you trust what you see or what you hear?
- Integration vs Segregation: When should you combine information from different senses and when should you treat them as independent?

Right: Psychophysics experimental setup with eye tracker

## **Reinforcement learning**

We are concerned with investigating how autonomous agents, either simulated or robotic, can learn via interaction with their environment.

### **Extending learning algorithms**



#### **Multiagent systems**

We are investigating ways in which agents can learn to cooperate with each other to produce cohesive team behaviour: • Rewarding individuals for group behaviour (left) Learning to use actions based on inter-



- agent communication • Learning by imitation
- Assembling robots from autonomous modules

Left: Multiagent reinforcement learning for commercial computer games Right: teleo-reactive programming in an educational soccer game



**Identifying Group Interactions** 

• Aim to establish who is interacting with who and the nature of

• Identification of interaction class. Are these people: following, meeting, walking together, fighting etc.?



- Work is being undertaken to extend traditional reinforcement learning algorithms, applying them to a number of complex tasks and environments:
  - Dynamic segmentation of the state/action space • Active perception: including actions for sensing in
    - ambiguous worlds (left)
  - Internal motivations or 'drives' for learning

Left: perceptual aliasing in a gridworld environment

