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Institute of Perception,
Action and Behaviour

Shared Autonomy for Robotics

Closing the Loop

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www.edinburgh-robotics.org



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ROBOTICS



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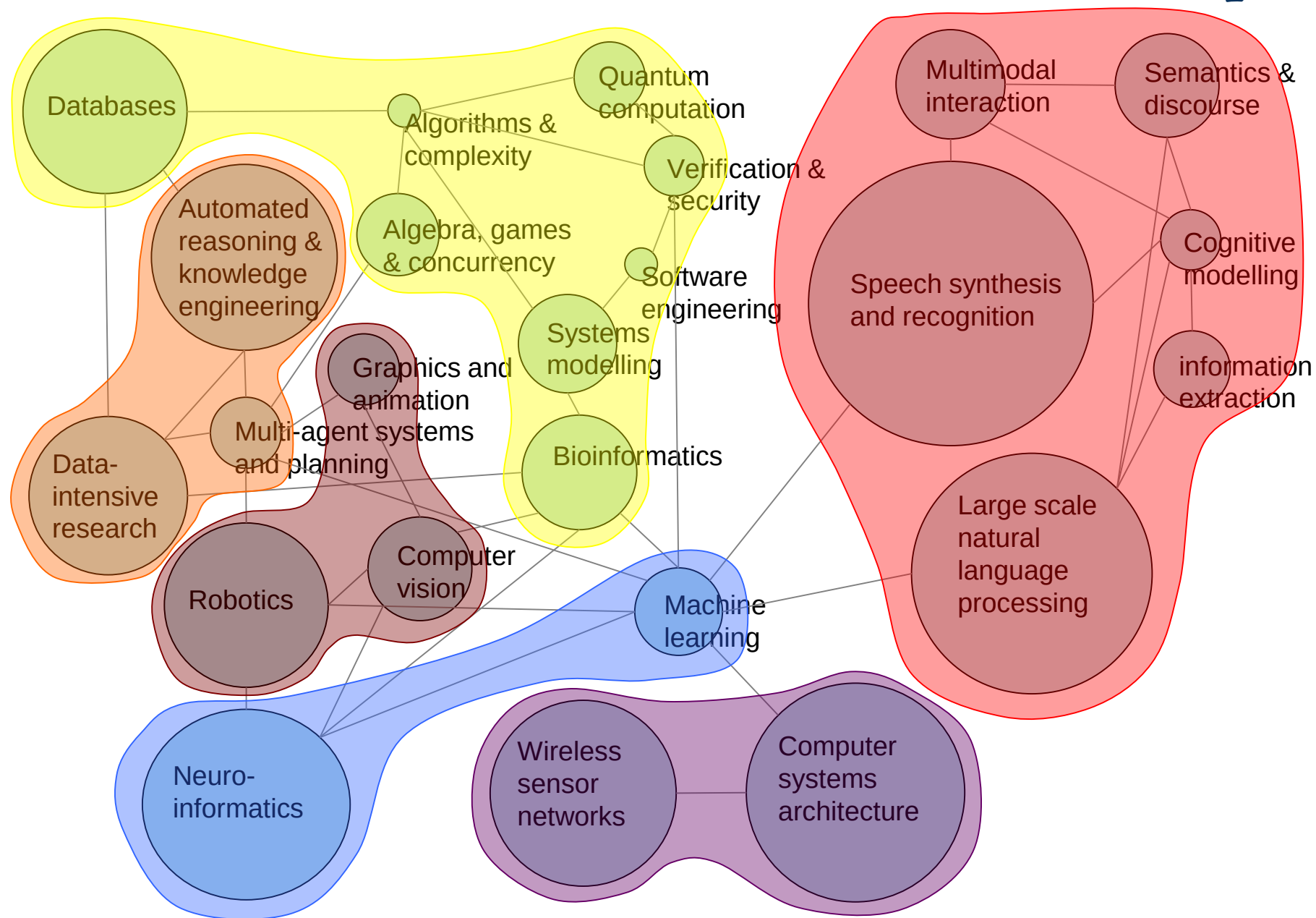
ROBOTICS AND COMPUTER VISION

Institute of Perception, Action and Behaviour (IPAB)

Director: Sethu Vijayakumar

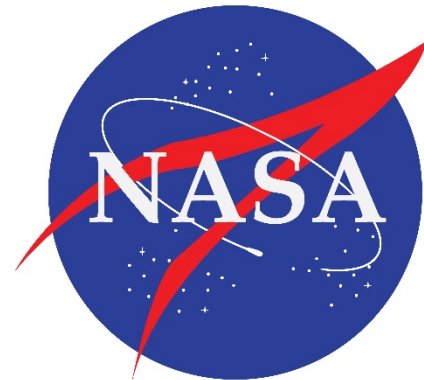


Our research landscape



Valkyrie Humanoid

Flagship Robotarium Platform:
UoE-NASA Valkyrie Humanoid
UK Hub for Humanoids Research
Space Robotics Challenge



NASA Space Robotics Challenge



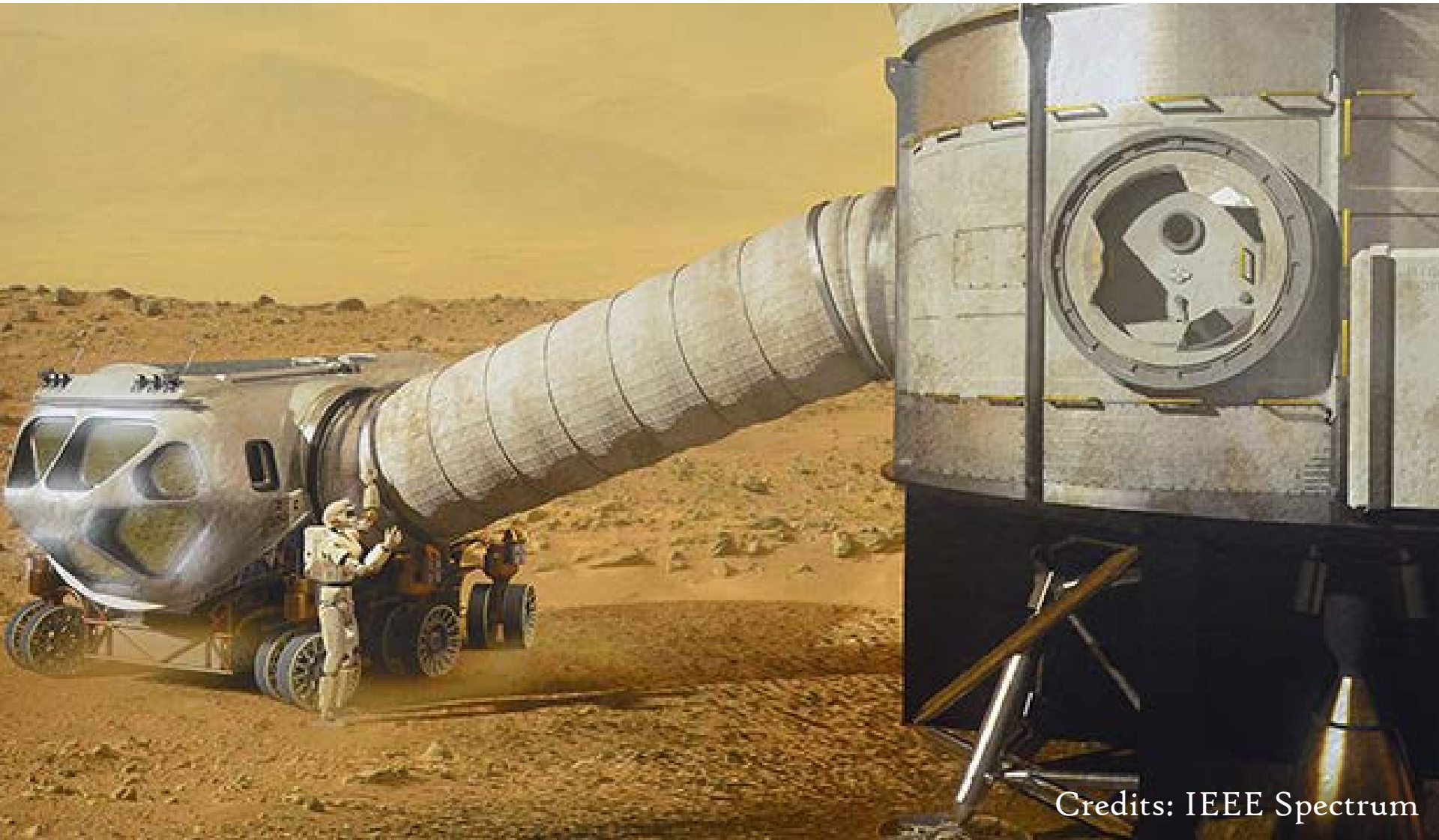
Credits: IEEE Spectrum

NASA Space Robotics Challenge



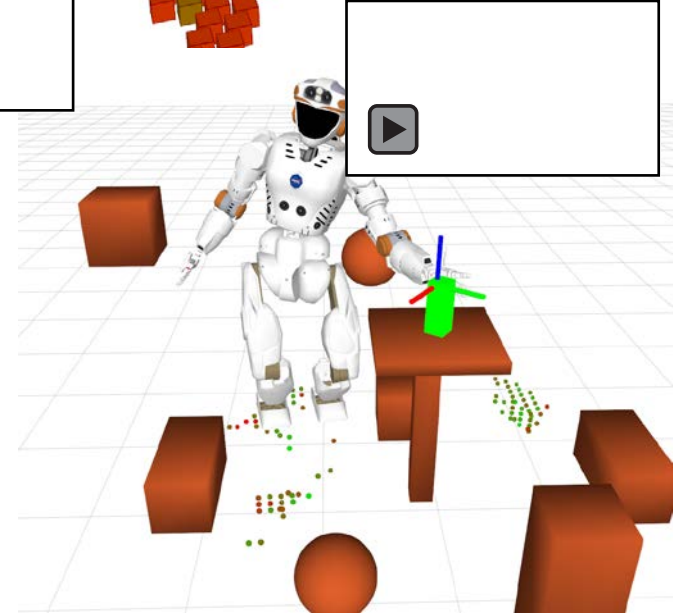
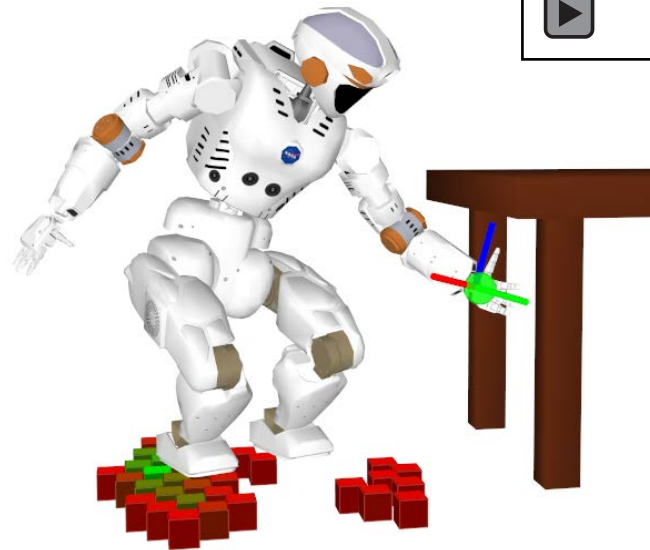
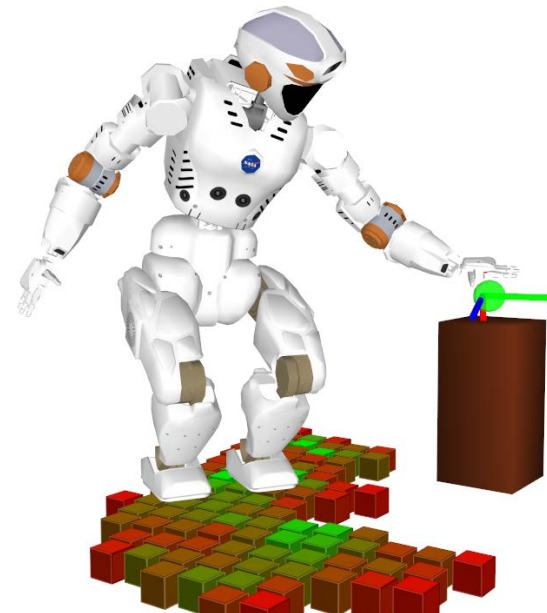
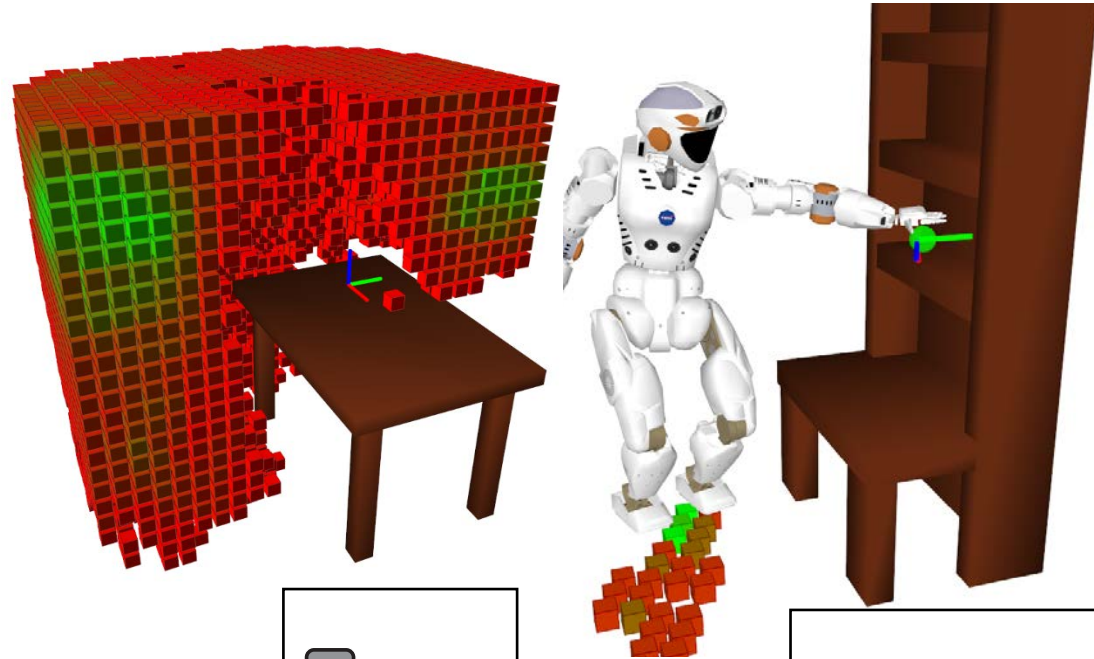
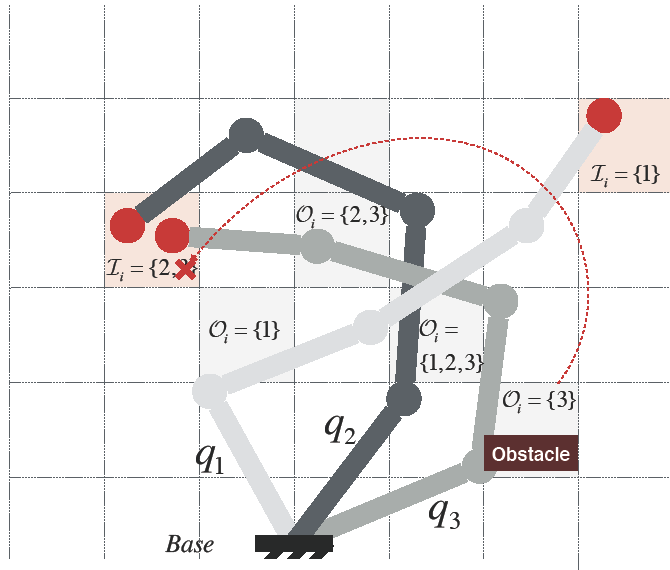
Credits: IEEE Spectrum

NASA Space Robotics Challenge

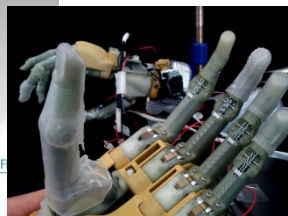


Credits: IEEE Spectrum

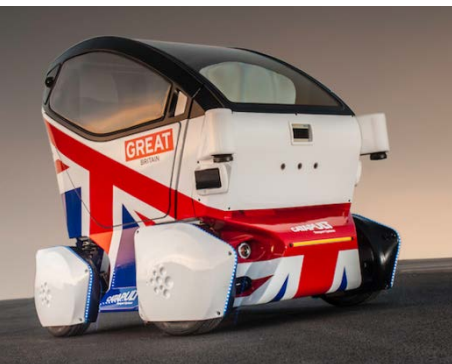
Advanced Whole Body Motion Planning



Robots That Interact



Prosthetics, Exoskeletons



Self Driving Cars



Field Robots (Marine)



Medical Robotics



Service Robots



Industrial/ Manufacturing

Full-autonomy challenges due to:

1. Close **interaction** with **multiple objects**
2. Noisy **sensing** with ambiguity
3. Hard to model **dynamics**
4. Guarantees for **safe operations**
5. Highly **constrained** environment
6. Un-modelled **user intentions**



Humanoids



Nuclear
Decommissioning

1: Understanding and Representing the world around you

(Tracking and Localisation)



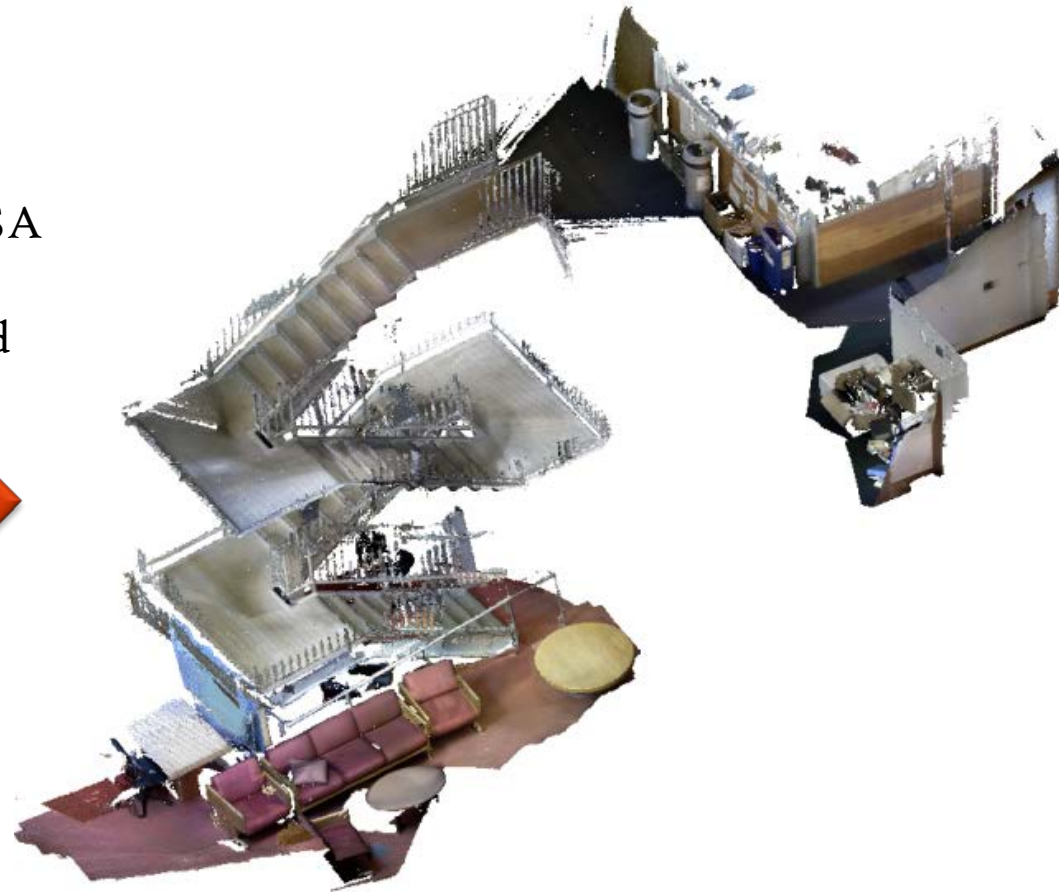
Toyota's Human Support Robot (HSR)

1: Understanding and Representing the world around you

State Estimation: **unified percept** of the world



UoE-NASA
Valkyrie
Humanoid
Platform



1: Understanding and Representing the world around you

(Closed Loop Sensing and Manipulation)

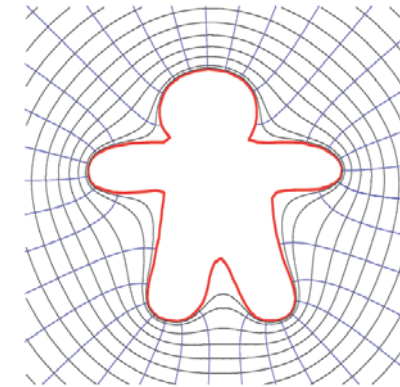
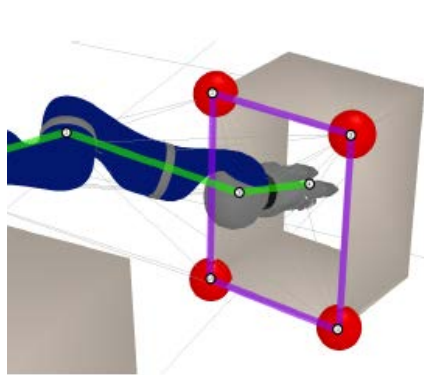


2: Understanding and Representing the world around you

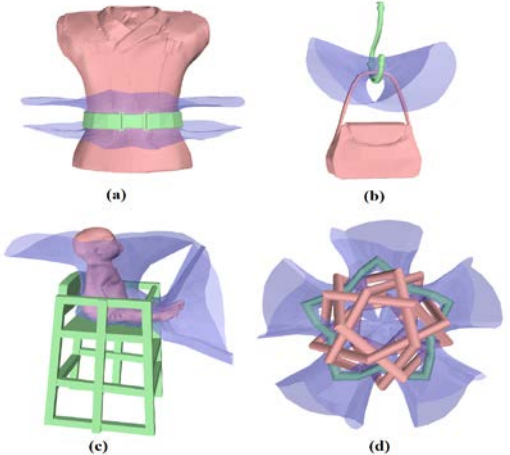
Bridging Representations



Interaction Mesh based Relational Descriptors



Harmonic Electric Fields



Relational tangent planes

- Interaction with dynamic, articulated and flexible bodies
- Departure from purely metric spaces -- focus on **relational metrics** between active robot parts and objects/environment
- Enables use of **simple motion priors** to express complex motion

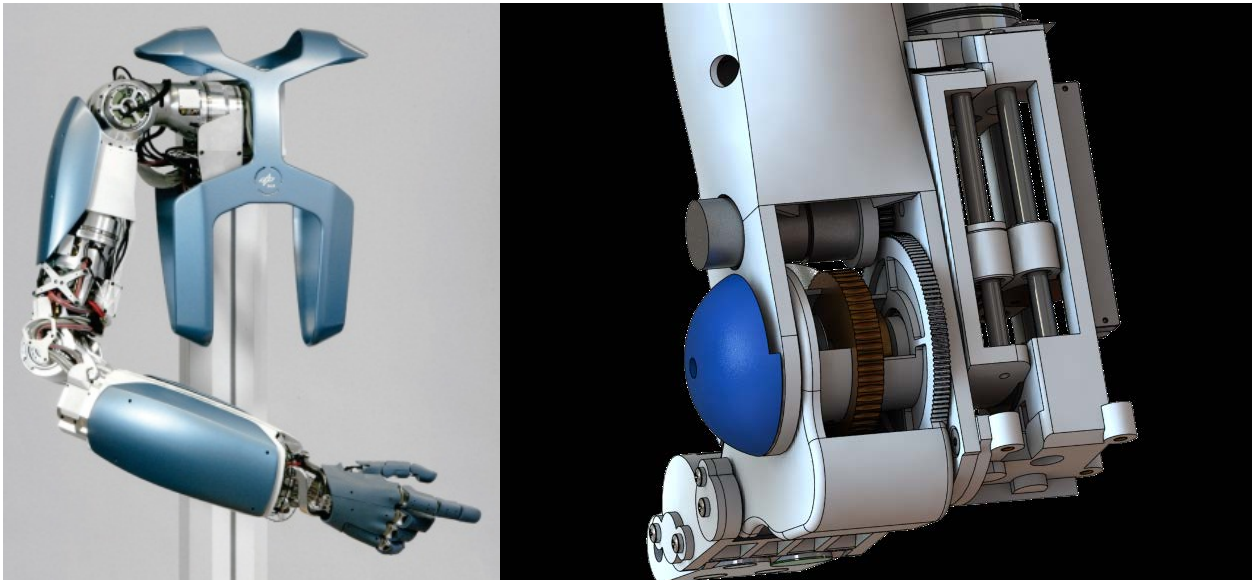
Robots for Confined Spaces



Courtesy: OC Robotics Ltd.

3:Dealing with Uncertainty

- through **Compliant Actuation**



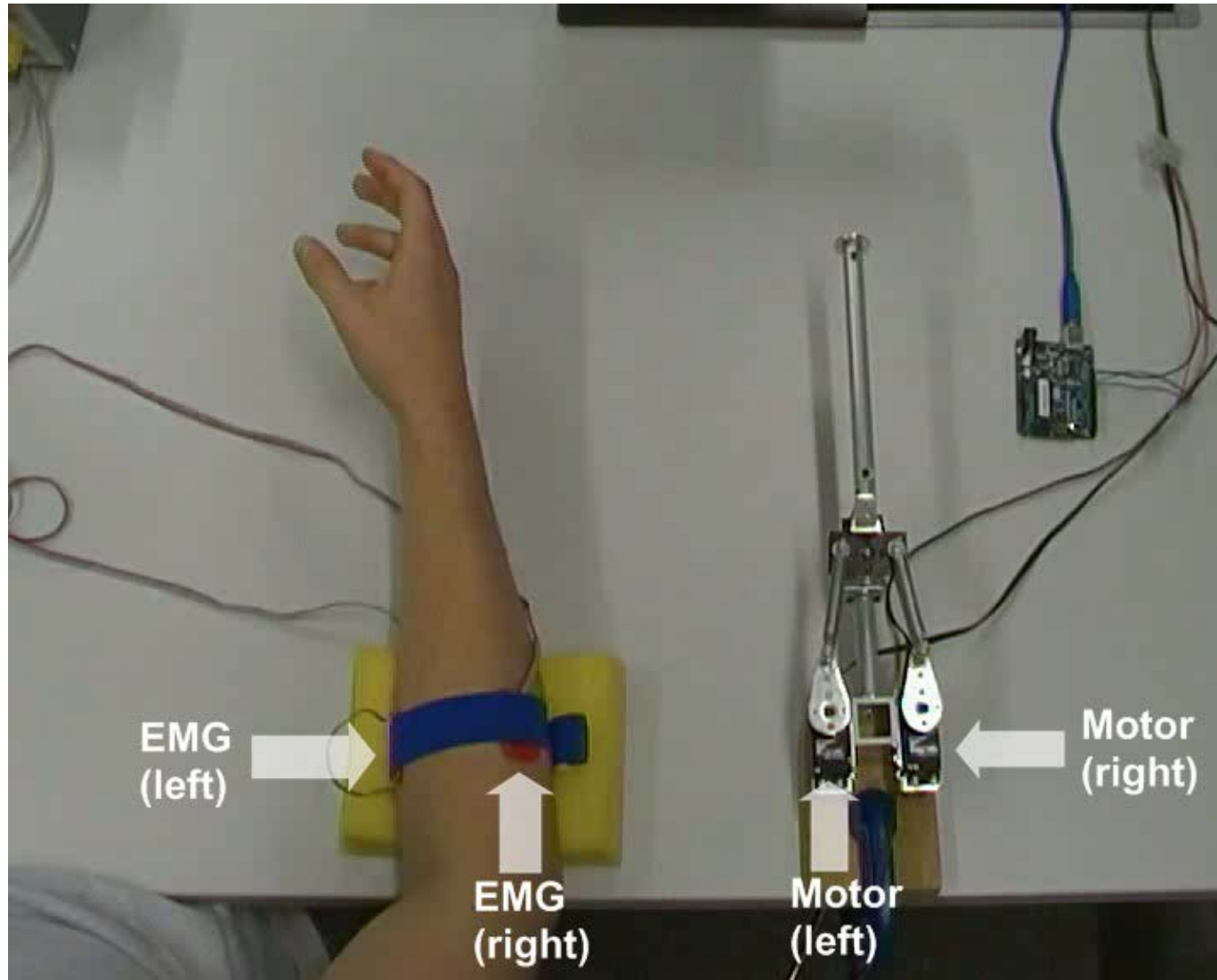
HASy (DLR); BLUE and miniBLUE compliant bipeds @ University of Edinburgh



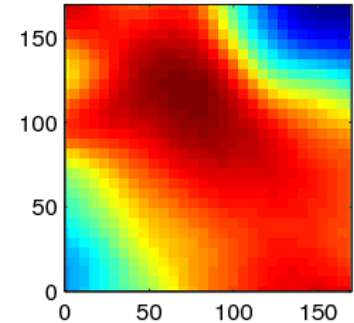
FED160710
T.07
Gillette Fusion Sensitive Skin
British Skin Foundation Shoot

Braun, Vijayakumar, et. al., Robots Driven by Compliant Actuators: Optimal Control under Actuation Constraints, *IEEE T-RO*, 29(5) (2013). [IEEE Transactions on Robotics Best Paper Award]

Variable Stiffness Actuation

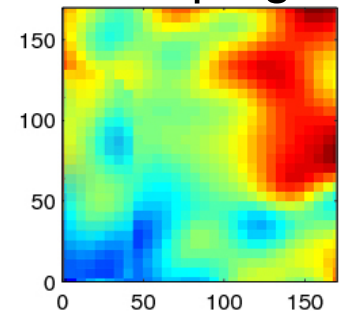


Stiffness



+

Damping



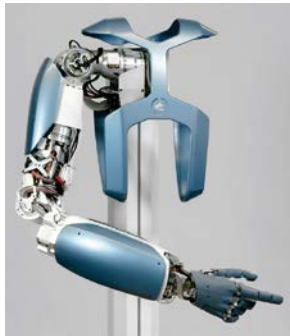
Impedance

Compliant Actuators

VARIABLE JOINT STIFFNESS



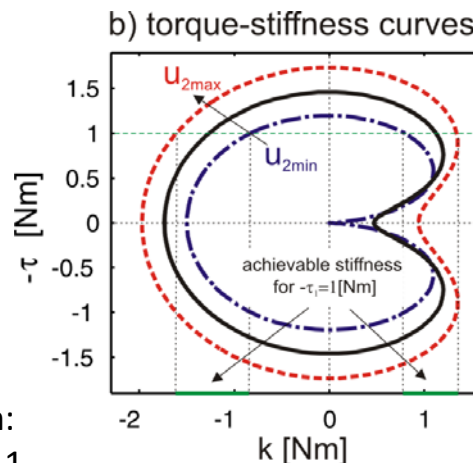
MACCEPA:
Van Ham et.al, 2007



DLR Hand Arm System:
Greibenstein et.al., 2011

$$\tau = \tau(\mathbf{q}, \mathbf{u})$$

$$\mathbf{K} = \mathbf{K}(\mathbf{q}, \mathbf{u})$$



Torque/Stiffness Opt.

Model of the system dynamics:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}) \quad \mathbf{u} \in \Omega$$

Control objective:

$$J = -d + w \frac{1}{2} \int_0^T \|\mathbf{F}\|^2 dt \rightarrow \min.$$

Optimal control solution:

$$\mathbf{u}(t, \mathbf{x}) = \mathbf{u}^*(t) + \mathbf{L}^*(t)(\mathbf{x} - \mathbf{x}^*(t))$$

iLQG: Li & Todorov 2007

DDP: Jacobson & Mayne 1970

Highly **dynamic** tasks, explosive movements



Optimising and Planning with Redundancy: **Stiffness** and **Movement** Parameters
Scale to High Dimensional Problems

Multi-phase Movement Optimization

- Task encoding of movement with multi-phases

$$J = \underbrace{\phi(\mathbf{x}(T_f))}_{(3)} + \sum_{j=1}^K \underbrace{\psi^j(\mathbf{x}(T_j^-))}_{(2)} + \int_{T_0}^{\overline{T_f}} \underbrace{h(\mathbf{x}, \mathbf{u})}_{(1)} dt$$

Terminal cost Via-point cost Running cost

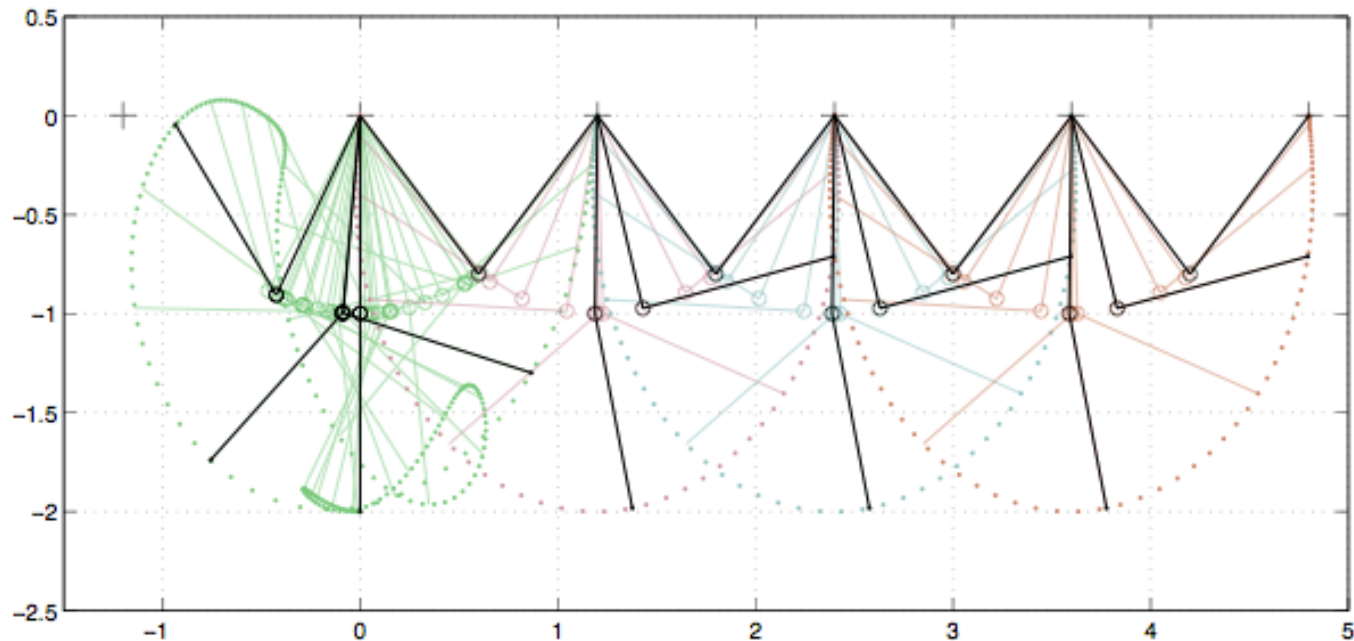
- cf. individual cost J_i for each phase $T_{j-1} \leq t < T_j$
- total cost by sequential optimization could be suboptimal

Optimization problem

- (1) optimal feedback control law $\mathbf{u} = \mathbf{u}(\mathbf{x}, t)$ to minimize J
- (2) switching instances $T_1, \dots, T_k$
- (3) final time (total movement duration) T_f

Optimized Brachiating Manoeuvre

Swing-up and locomotion



Additionally optimize for Movement Time

4: Learning to Predict and Adapt

- Predicting Consequences
- Predicting Task Goals and Intentions

Learning the Internal Dynamics

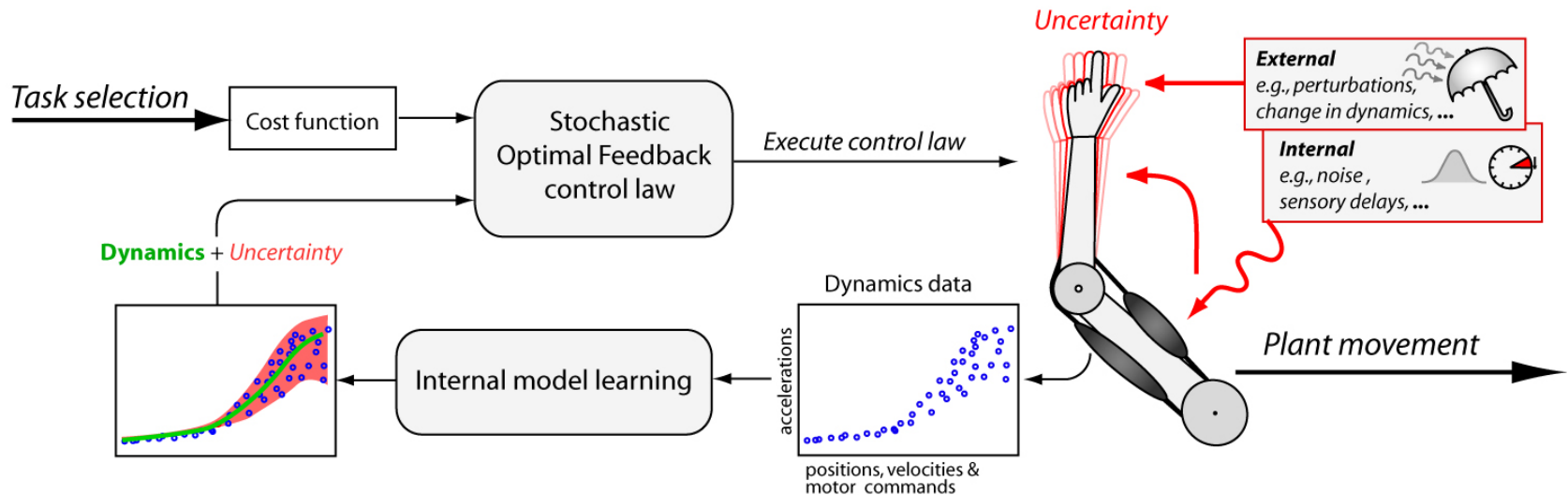
Learning the Task Dynamics



Stefan Klanke, Sethu Vijayakumar and Stefan Schaal, A Library for Locally Weighted Projection Regression, *Journal of Machine Learning Research (JMLR)*, vol. 9. pp. 623--626 (2008).

<http://www.ipab.inf.ed.ac.uk/slmc/software/lwpr>

On-the-fly **adaptation** at Any Scale



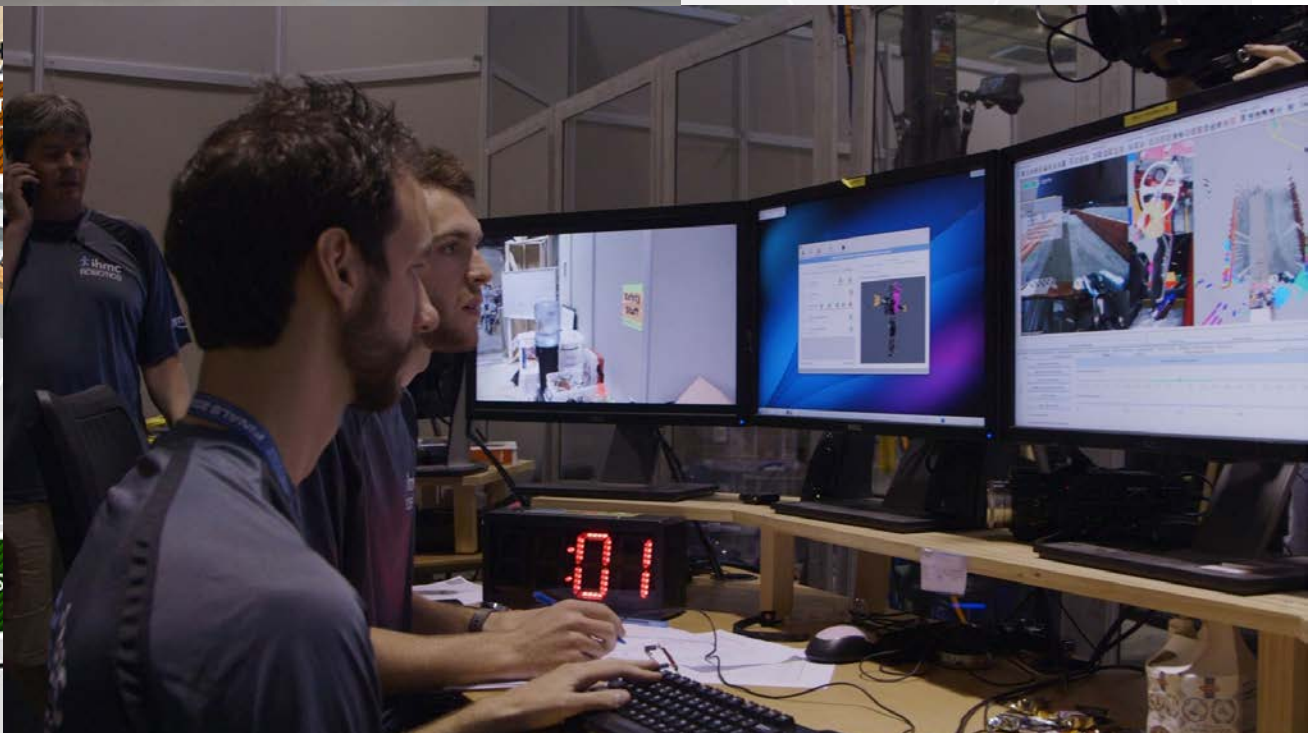
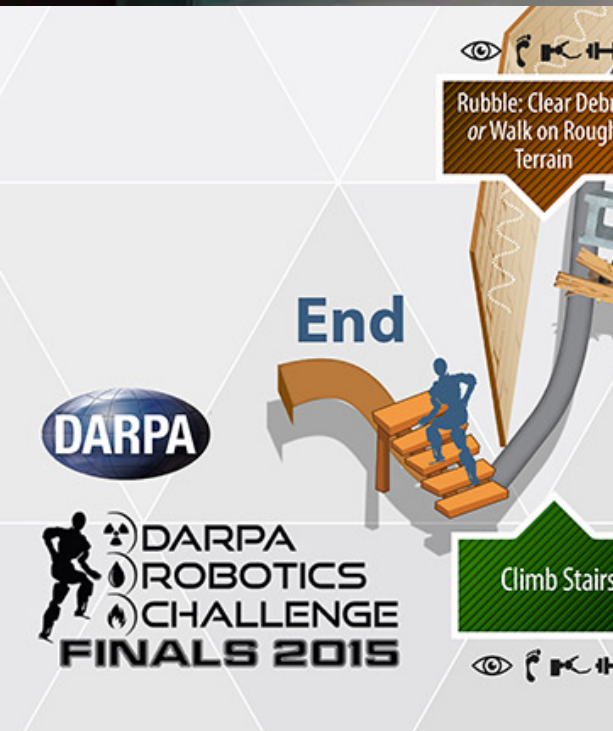
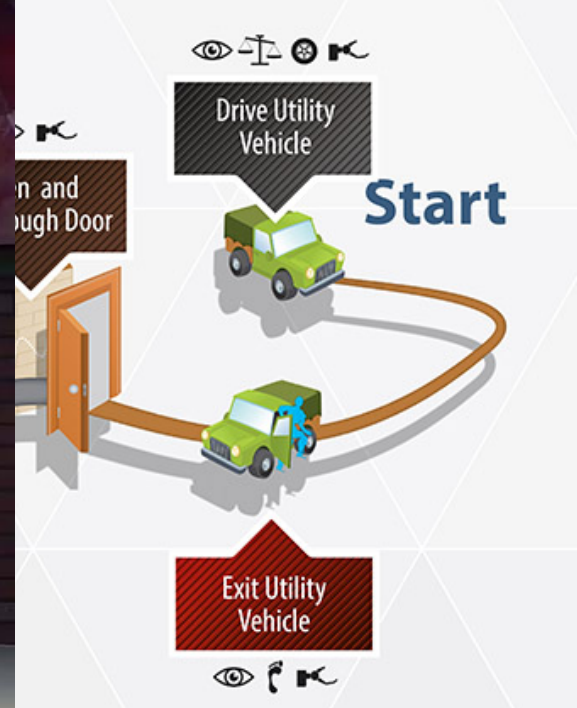
- Fast dynamics online learning for adaptation
- Fast (re) planning methods that incorporate dynamics adaptation
- Efficient Any Scale (embedded, cloud, tethered) implementation

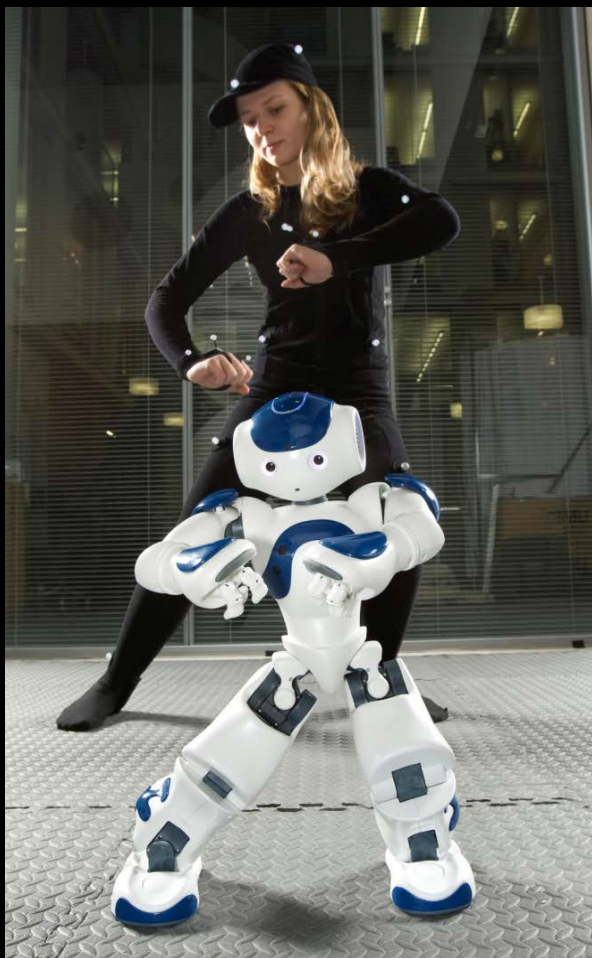
Putting it all together:

Adaptive, Human-in-the Loop Behaviour

This capability is crucial for **safe, yet precise** human robot interactions as well as applications as diverse as **wearable exoskeletons**.



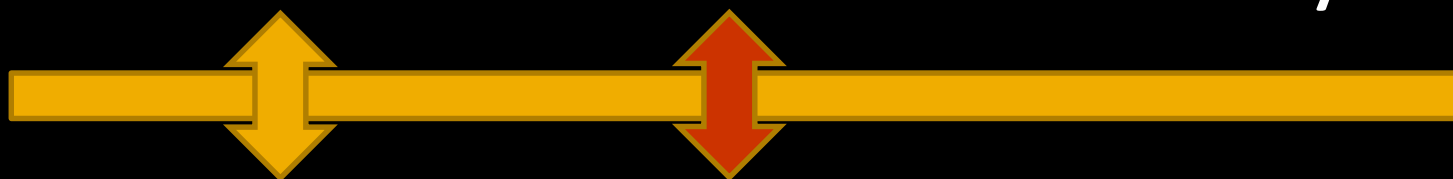




Teleoperation



Autonomy



Shared Autonomy

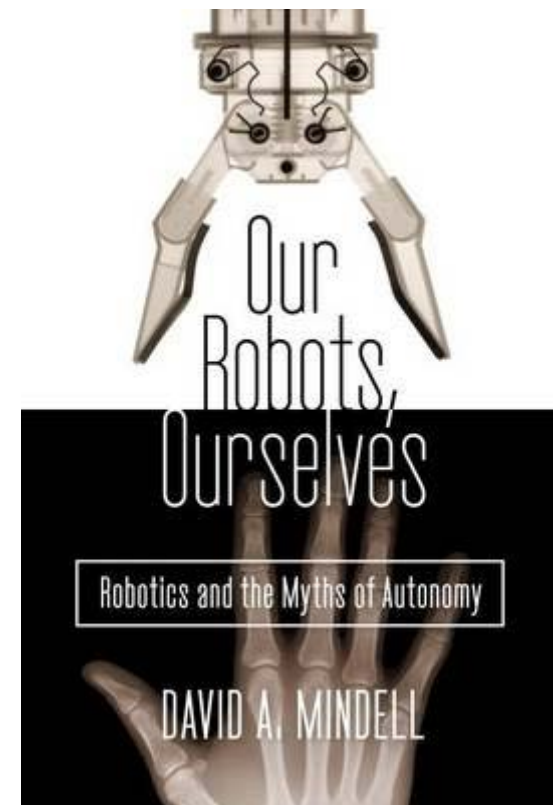
Shared Autonomy

“Full autonomy is a 20th century rhetoric”

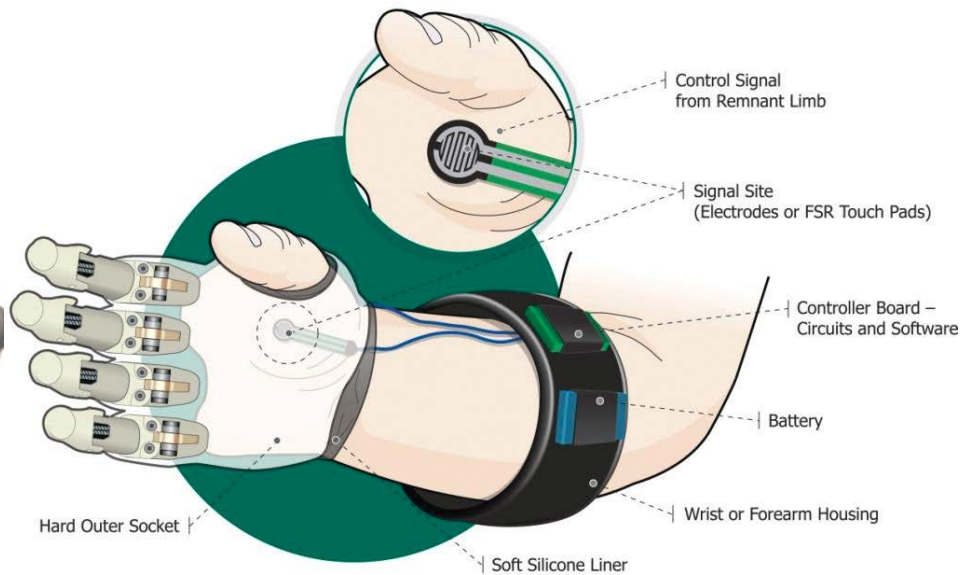
“The real frontier is collaboration, which includes autonomy but **different levels of autonomy at different moments** under the control of a human operator.”

“Such systems should have the ability to turn on autonomy when it can be helpful. Autonomy can reduce human workload and fatigue, but humans should still be present in the loop.”

- IEEE Spectrum, David Mindell (MIT)



Shared Autonomy: An Example





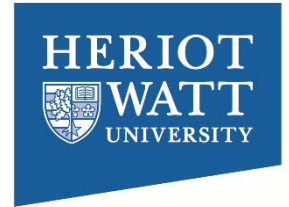
EPSRC CDT-RAS

The EPSRC Center for Doctoral Training in **Robotics & Autonomous Systems**

- **Multidisciplinary ecosystem** – 65 PhD graduates over 8.5 years, 50 PIs across Engineering and Informatics disciplines
Control, actuation, Machine learning, AI, neural computation, photonics, decision making, language cognition, human-robot interaction, image processing, manufacture research, ocean systems ...
- **Technical focus** – ‘Interaction’ in Robotic Systems
Environment: Multi-Robot: People: Self: Enablers
- **‘Innovation Ready’ postgraduates**
Populate the innovation pipeline. Create new businesses and models.
- **Cross sector exploitation**
Offshore energy, search & rescue, medical, rehabilitation, ageing, manufacturing, space, nuclear, defence, aerospace, environment monitoring, transport, education, entertainment ..
- **Total Award Value (> £14M)**: CDT £7M, Robotarium £7.1M
38 company sponsors, £2M cash, £6.5M in-kind (so far ..)
Schlumberger, Baker Hughes,, Renishaw, Honda, Network Rail, Selex, Thales, BAe, BP, Pelamis, Aquamarine Power, SciSys, Shadow Robot, SeeByte, Touch Bionics, Marza, OC Robotics, KUKA, Dyson, Agilent ...



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ROBOTARIUM

A National UK Facility for Research into the Interactions amongst
Robots, Environments, People and Autonomous Systems

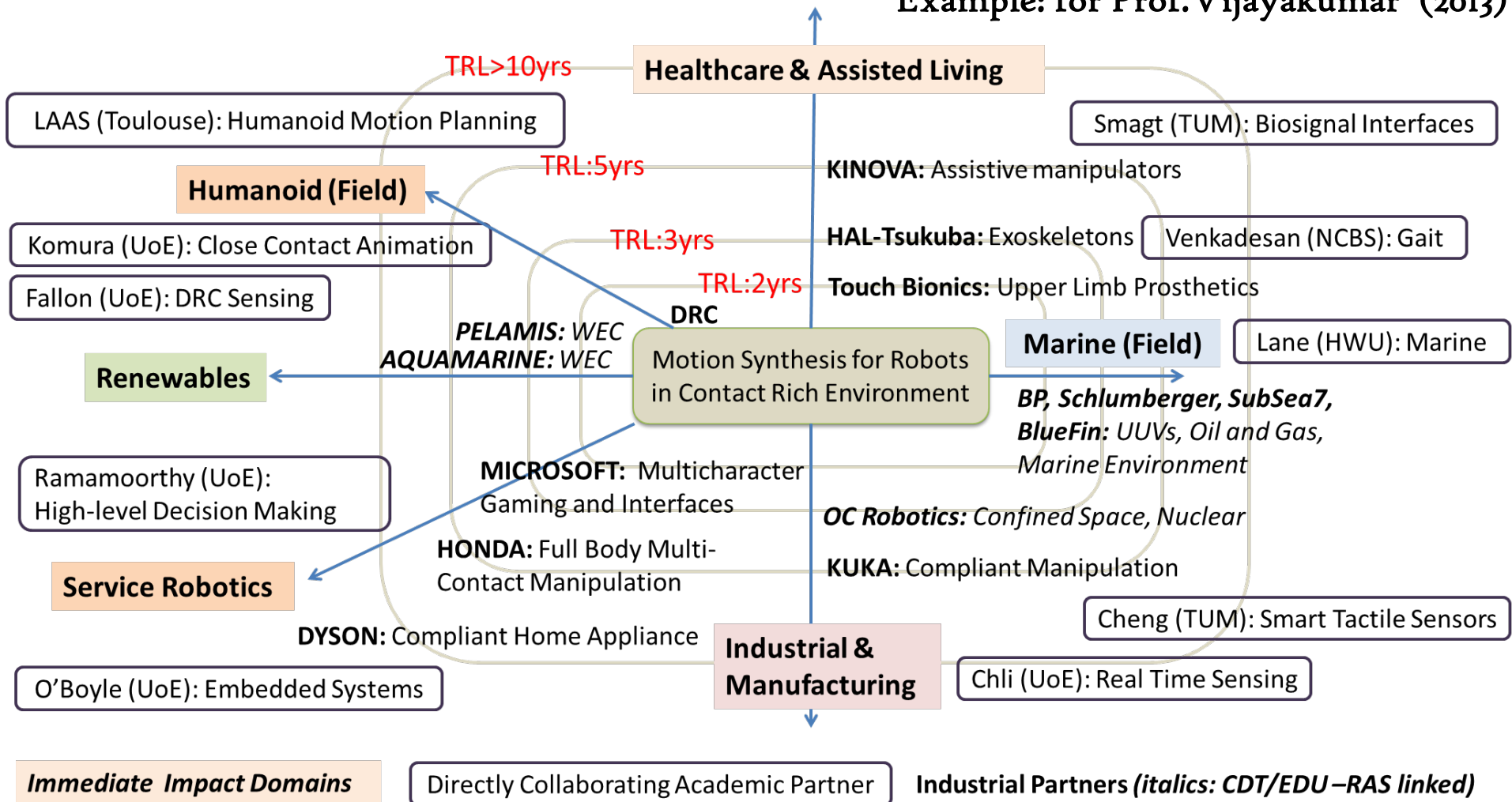


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EPSRC

Translation and Impact

Example: for Prof. Vijayakumar (2013)



- Translation through **Industrial & Scientific** Collaborations and Skilled **People**

“we tend to overestimate technology
in the short term and
underestimate technology
in the long term”

Arthur C. Clarke

interactive

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