

ANITI

ARTIFICIAL & NATURAL INTELLIGENCE
TOULOUSE INSTITUTE



(Artificial and Natural) Motion

Nicolas Mansard, Philippe Souères, Olivier Stasse

- **General presentation**
- **Some results**
 - Scientific results
 - Related works
 - Planned PhD / post doc proposals
- **Interaction with other chairs / industrial**

Chair members



Olivier Stasse



Nicolas Mansard



Philippe Souères

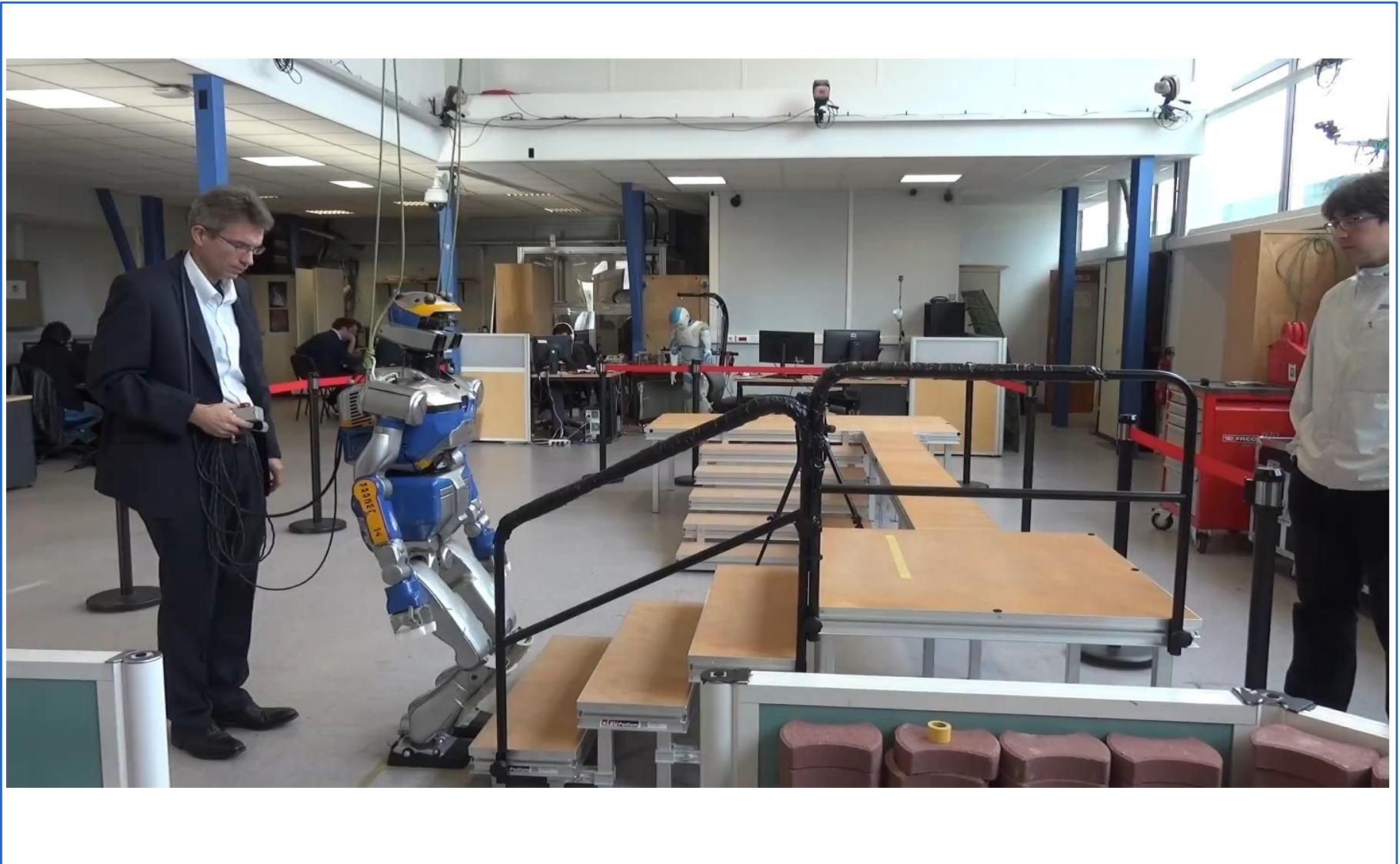


Pyrene & HRP-2

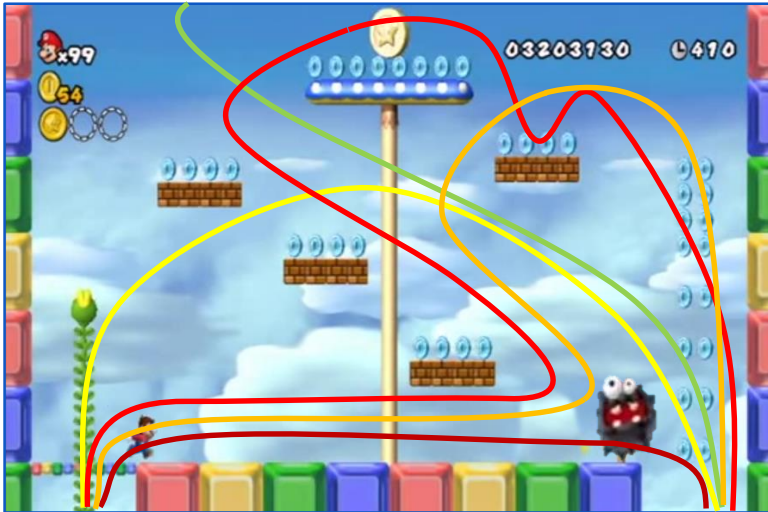
Chair members



Gepetto @ LAAS



Optimal control – as a universal toolbox



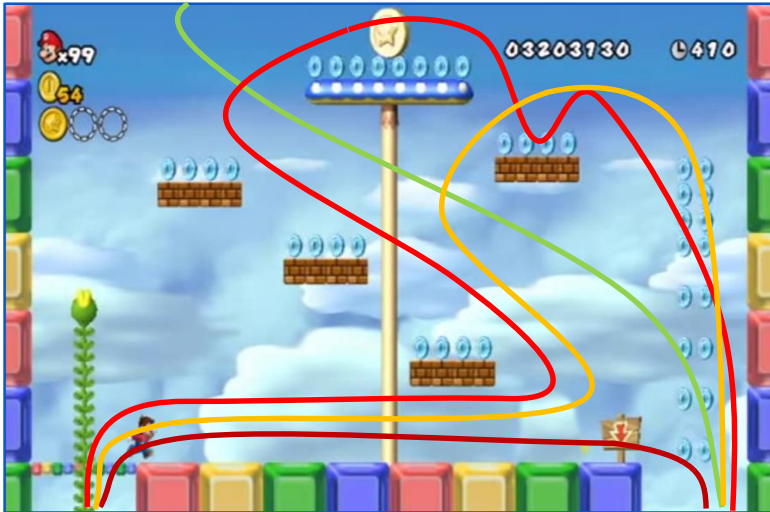
Dynamics

$$\dot{x} = f(x, u)$$

State
(e.g. configuration
+ velocity)

Control
(e.g. motor
torques)

Optimal control – as a universal toolbox



Dynamics

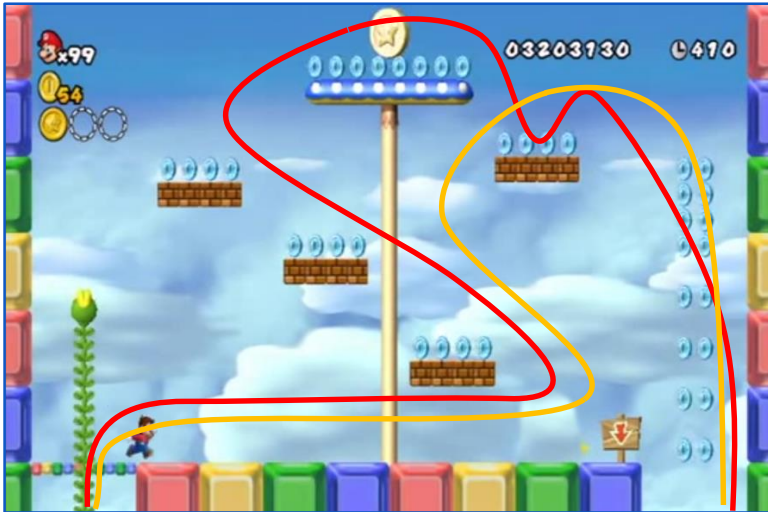
$$\dot{x} = f(x, u)$$

Constraints

$$x(0) = x_0$$

$$h(x, u) \geq 0$$

Optimal control – as a universal toolbox



Dynamics

$$\dot{x} = f(x, u)$$

Constraints

$$x(0) = x_0$$

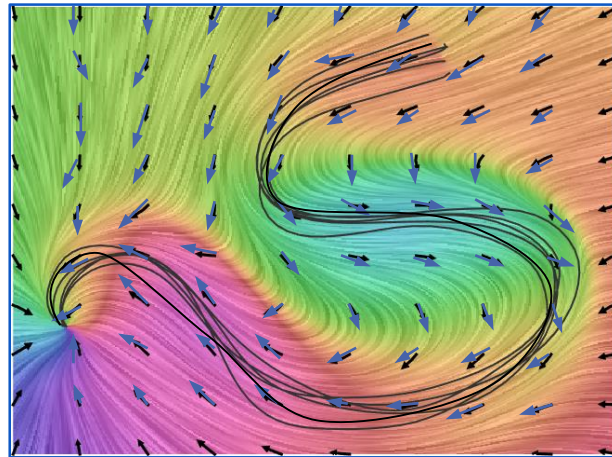
$$h(x, u) \geq 0$$

Cost / reward

$$\int_0^T l(x(t), u(t)) dt$$

$$\min_{\{x_t\}, \{u_t\}} \int_0^T l_1(x_t, u_t) dt \quad \text{AND} \quad l_2(x_T) \quad \dots \quad \text{AND} \quad \int_0^T l_p(x_t, u_t) dt$$

so that $\forall t, \dot{x}(t) = f(x(t), u(t))$



Optimizing a trajectory

- $U: t \rightarrow u(t)$
- Motion planning

Optimizing a policy

- $\Pi: x \rightarrow u = \Pi(x)$
- Reinforcement learning

Model versus data in legged robotics



Boston Dynamics

- Feedback linearization
- Expert tuning

- Few trials and errors



DeepMind

- Policy learning
- Bullet-proof simulator

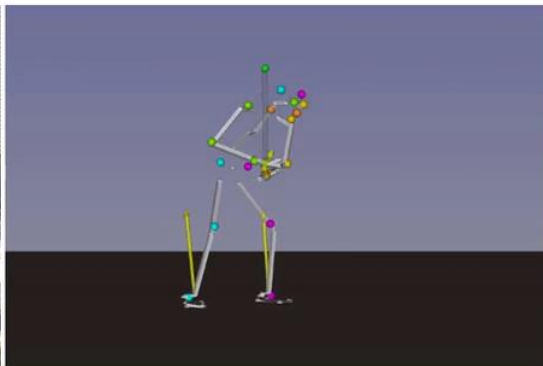
- Tons of data

Data-driven strategies ... some examples

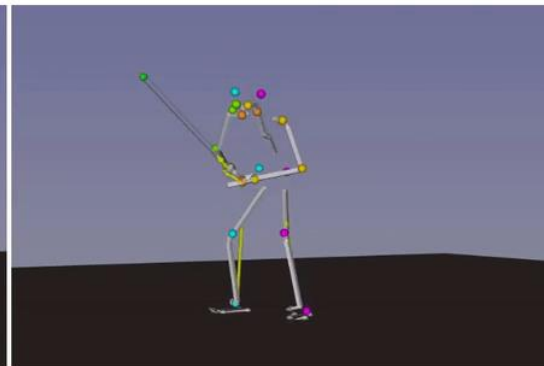
Hammer



Input video



Output motion and forces



Output from another viewpoint

CVPR'19, best-paper finalist

Computer vision (with Willow@Inria/Prairie)

- ≡ How to detect an axe?
- ≡ Where to find a database of axe images?

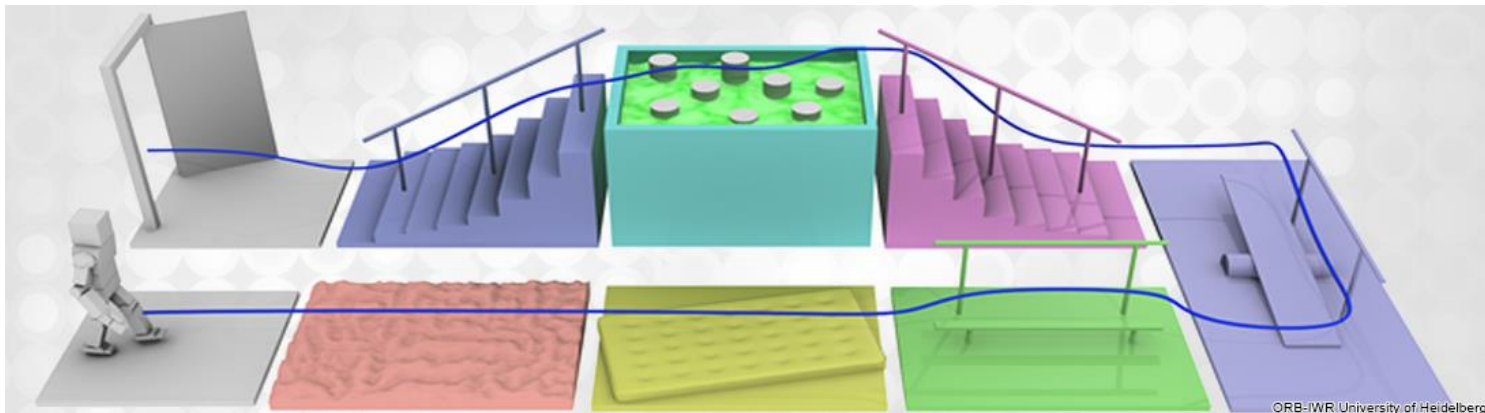
Understand how **data** and **models** can work together

Theoretical side

- Joint policy (RL = data) and trajectory (planning=model) optimization
- Realistic constrained optimization (real-time, robust, fast, etc)

Practical side

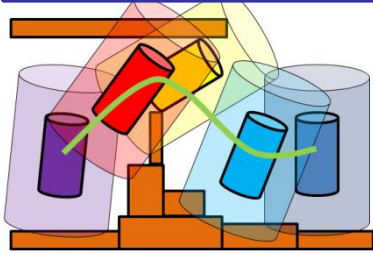
- Solve the locomotion problem
- Generalize it to other problems in robotics (manipulation first)



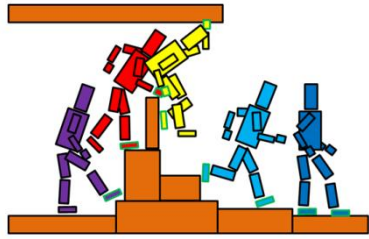
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Locomotion: model-based heuristics

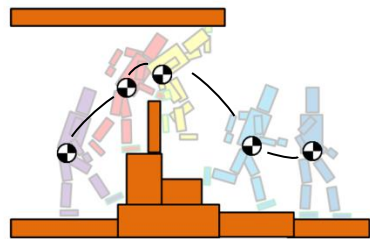
Reachability planner



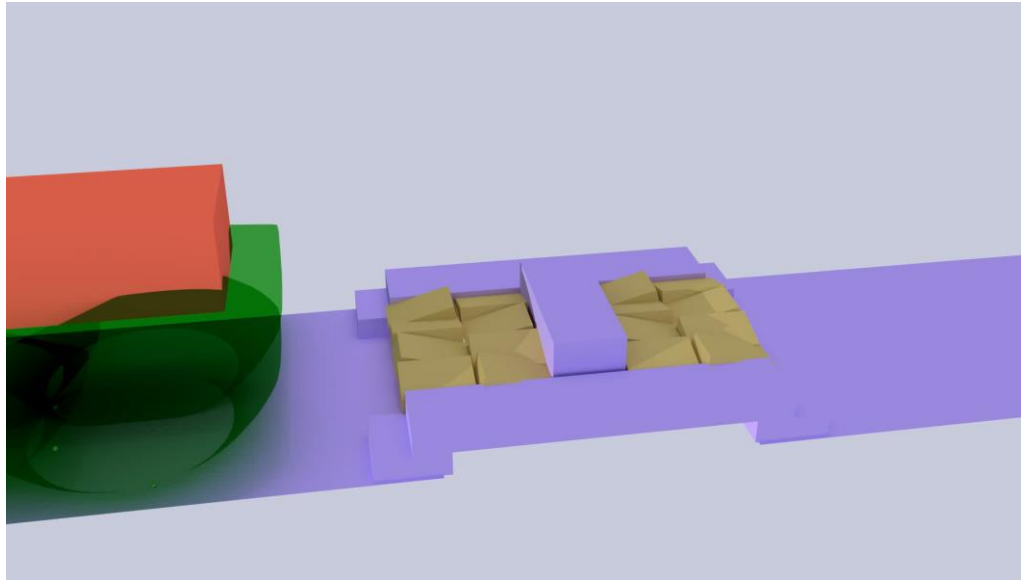
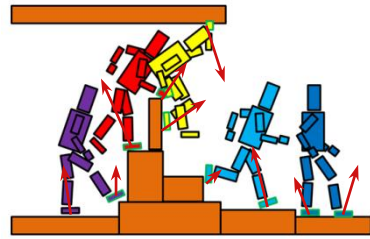
Sequence of contacts



3D Pattern generator

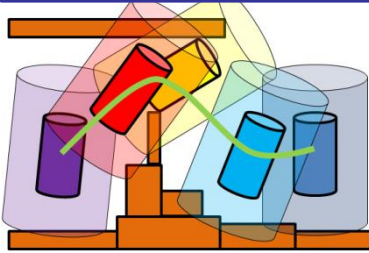


Whole-body force control

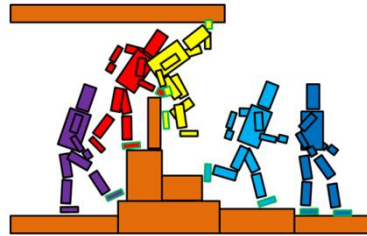


Locomotion: model-based heuristics

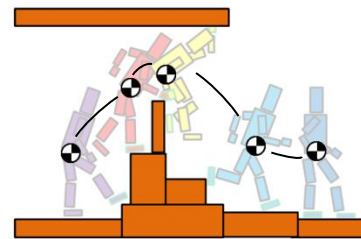
Reachability
planner



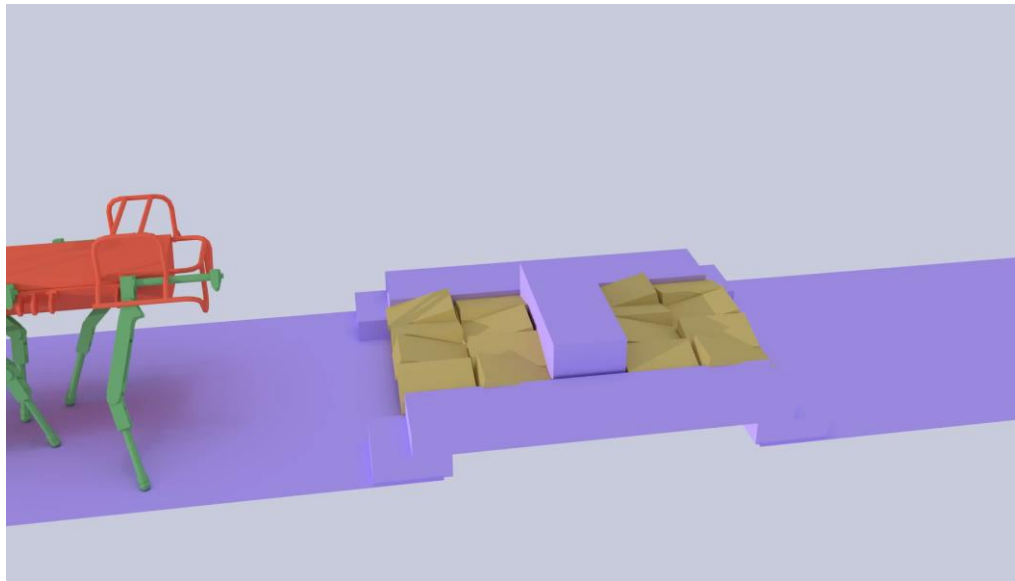
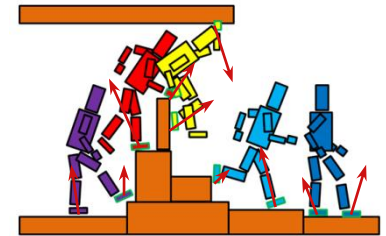
Sequence of
contacts



3D Pattern
generator

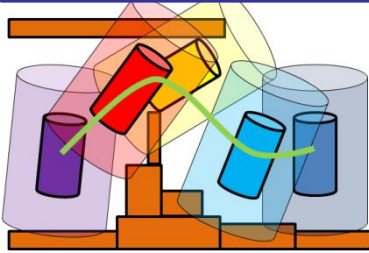


Whole-body
force control

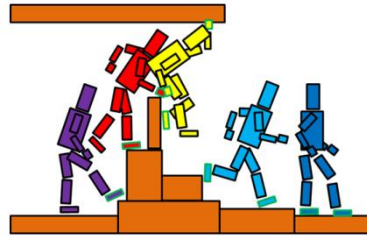


Locomotion: model-based heuristics

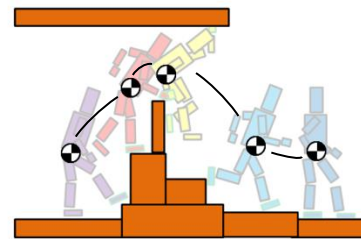
Reachability
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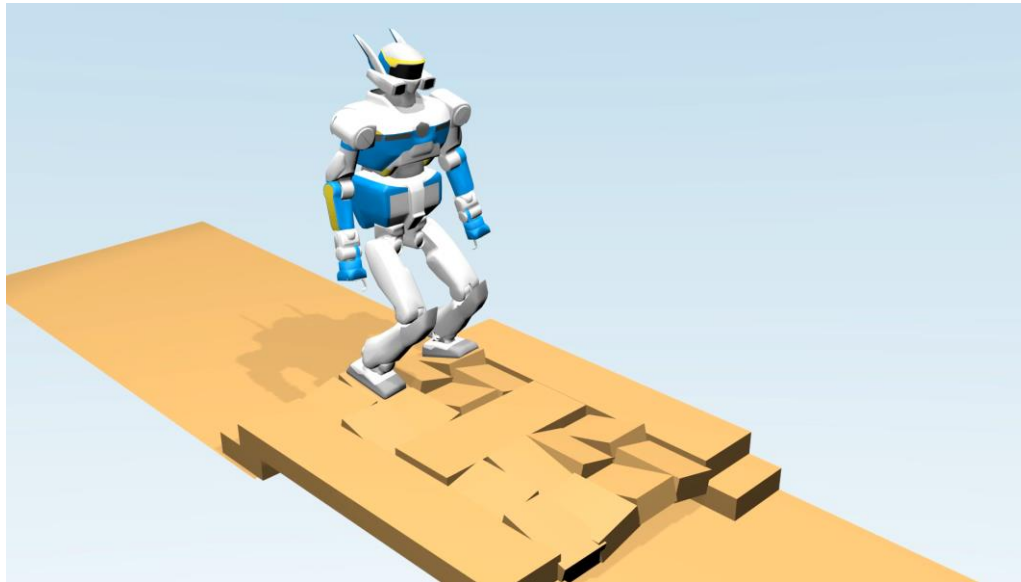
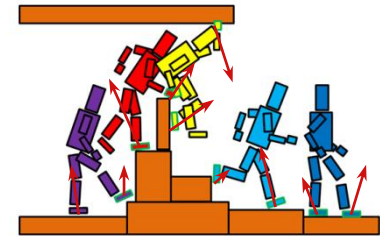
Sequence of
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3D Pattern
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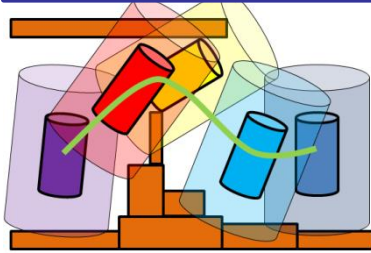


Whole-body
force control

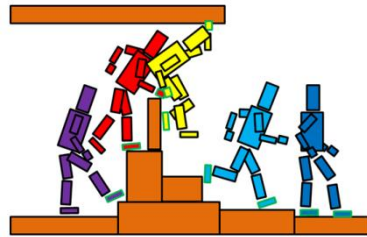


Locomotion: model-based heuristics

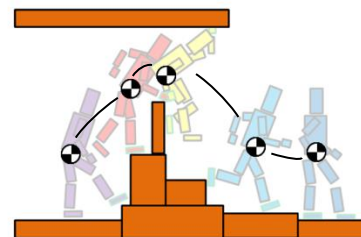
Reachability
planner



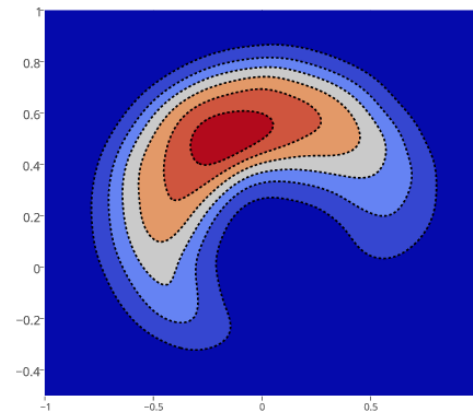
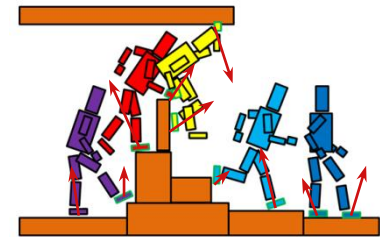
Sequence of
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3D Pattern
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Whole-body
force control



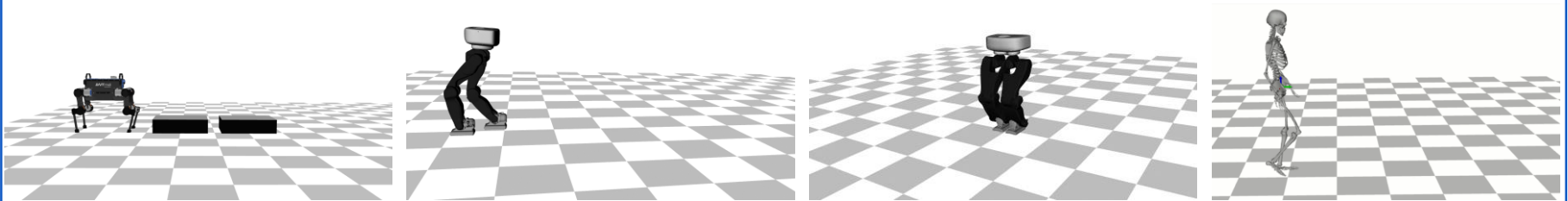
Data-based GM model

Project with H2020 Memmo and Prairie

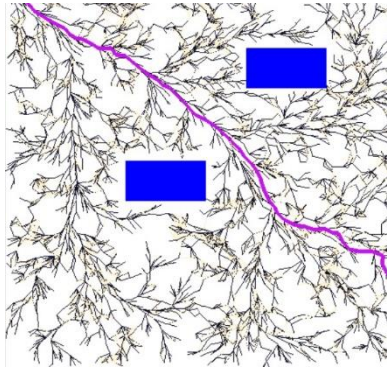
$$\min_{\substack{X=(Q,\dot{Q}), \\ U=\tau}} \int_0^T l(x_t, u_t) dt$$

so that $\forall t, \dot{x}(t) = \underbrace{f(x(t), u(t))}_{\text{robot dynamics}}, \underbrace{g(x(t), u(t)) = 0}_{\text{robot dynamics}}$

- Differential dynamic programming (sparse SQP with multiple shooting)
 - $\text{Dim}(x) = 50, \text{dim}(u) = 20, T = 200 \dots \text{total} = 14,000$ variables
 - One descent step = 7 ms
- Introduce additional hard constraints with augmented Lagrangian

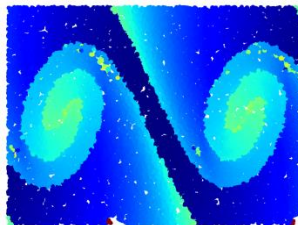


Model-based reinforcement learning



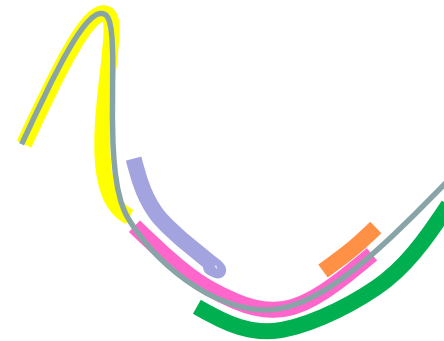
**Kino-dynamic
Probabilistic Roadmap**
30-50 states, dense connect

↑ Roadmap
extension



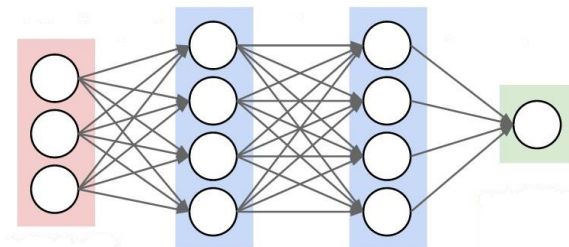
HJB approximation
*Value function as metric
Policy function as warm-start*

→
→
→
→
→
Sampling



Dataset of subtrajectories
10-100k items

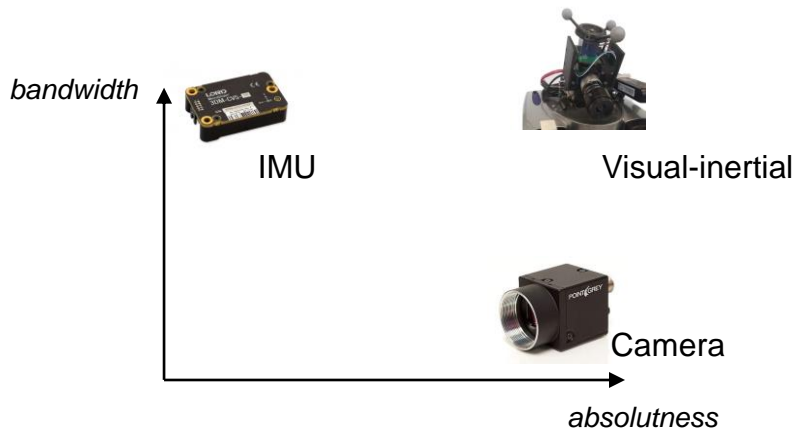
↓ Regression
(stoch.grad.)



Neural network
2x512 hidden units

← Query

□ Two complementary sensors



□ Optimal estimation (max likelihood)

- Camera: artificial (or natural) landmarks
- IMU: pre-integration in Lie groups
- Optimization: off-constraint stochastic optimization

- ❑ PhD thesis in constrained optimization ... Ewen Dantec
- ❑ PhD thesis in model-based reinforcement learning ... Amit Parag
- ❑ Looking for a post-doc
 - ❑ Expertise in robotics or **numerical optimization** welcome

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- ❑ Industrial collaboration to be emphasized in ANITI
 - ❑ AIRBUS joint lab ROB4FAM (future of aerospace manufacturing)
 - ❑ Naver Labs: co-supervision of a PhD
 - ❑ Proposition of CIFRE by Continental Corp

- ❑ Academic collaboration
 - ❑ Coordinator of H2020 Memmo
Max-Planck Inst., Univ. Edinburgh, IDIAP Swiss
 - ❑ Collaboration with 3IA
with PRAIRIE (Ponce, Laumond, Carpentier, Schmidt)
with MIAI (Naver Labs, Schmidt)



- ❑ In (continuous) constrained optimization
 - ❑ Jérôme Bolte, Edouard Pauwel, Marc Teboulle, JB Lasserre
 - ❑ GANs? Jérôme Renault

- ❑ In data-driven reinforcement learning
 - ❑ Emmanuel Rachelson

- ❑ In computer vision ?

- ❑ Looking for data?
We have many experimental setup for you !

