### Safe Motion Control in physical Human-Robot Interaction

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Source: Robot & Frank





Source: MIT Media Lab



Interaction



Source: KUKA

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### **Physical Interaction**



- 1. Act compliant
- 2. Feel and understand contacts
- 3. Move safely





### Safe Real-Time Motion Planning







### Depth Sensors: From Stereo Vision to Kinect

**Stereo Vision** 





Time of Flight

Structured Light









Work of Andrea Maria Zanchettin and Paolo Rocco in Safe Motion Planning!

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C-Space Planning: Kulić & Croft 2006-2006

$$J = w_G f_G(d_G) + w_0 f_0(d_0) + w_D K \cdot DC$$

- "Extension" of (Nokata, Ikuta et al., 2002)
- C-space planner based on *best-first planning approach*, meaning following steepest decent of cost function (Latombe, 1991)
- Danger Criterion:

$$DC = \frac{I_S}{I_{\max}} f_{CoM}(D_{CoM})$$

- *Reflected inertia* based safety coefficient
- *Distance based* safety coefficient



### Optimization: Bicchi et. al. 2003-2005



 $\begin{cases} \min_T \int_0^T 1 \ dt \\ M_{\text{rot}} \ddot{x}_{\text{rot}} + K_{\text{transm}}(x_{\text{rot}} - x_{\text{link}}) = u \\ M_{\text{link}} \ddot{x}_{\text{link}} + K_{\text{transm}}(x_{\text{link}} - x_{\text{rot}}) = 0 \\ |\dot{x}_{\text{link}}| \le \nu_{\text{safe}}(K_{\text{transm}}) \\ |u| \le U_{\text{max}}, \end{cases}$ 

with initial and terminal conditions:









### **Real-Time Motion Planning**





### Potential Fields: Khatib 1986

#### The manipulator moves in a field of forces. The position to be reached is an attractive pole for the end effector and obstacles are repulsive surfaces for the manipulator parts.





### Potential Fields: Khatib 1986





### **Dynamical Systems for Collision Avoidance**

2<sup>nd</sup> order dynamical system:  $M\ddot{x}_d + K(x_d - x_g) + D\dot{x}_d = 0$ 

Potential Fields

$$M\ddot{x}_{d} + K(x_{d} - x_{g}) + D\dot{x}_{d} = \sum_{i} F_{i}(x_{o,i}, x_{d}, ...)$$

$$Task knowledge$$

$$M\ddot{x}_{d} + K(x_{d} - x_{g}) + D\dot{x}_{d} = \sum_{i} F_{i}(x_{o,i}, x_{d}, ...) + F_{task}$$

Potential Fields+Task feedforward, like DMP

$$M\ddot{x}_{d} + K(x_{d} - x_{g}) + D\dot{x}_{d} = \sum F_{i}(x_{o,i}, x_{d}, ...)$$

Variable Attractor Dynamics +  $M\ddot{x}_d + K(x_d - x_g) + D\dot{x}_d = \sum F_i(x_{o,i}, x_d, ...)$ Velocity scaling



### Variable Attractor Dynamics: Haddadin et al. 2009







### Variable Attractor Dynamics: Haddadin et al. 2009





### Dynamic Movement Primitives: Park et al. 2008

$$egin{array}{rll} au &=& \mathbf{K}(\mathbf{g}-\mathbf{x}) - \mathbf{D}\mathbf{v} - \mathbf{K}(\mathbf{g}-\mathbf{x}_0) heta + \mathbf{K}\mathbf{f}( heta) \ && + arphi(\mathbf{x},\mathbf{v}) \ && au\dot{\mathbf{x}} &=& \mathbf{v} \ . \end{array}$$

$$\varphi(\mathbf{x}, \mathbf{v}) = -\nabla_x U_{\text{dyn}}(\mathbf{x}, \mathbf{v})$$
$$= \lambda (-\cos\theta)^{\beta - 1} \frac{||\mathbf{v}||}{p} \left(\beta \nabla_x \cos\theta - \frac{\cos\theta}{p} \nabla_x p\right)$$

$$U_{\text{static}}(\mathbf{x}) = \left\{ \begin{array}{cc} \frac{\eta}{2} \left( \frac{1}{p(\mathbf{x})} - \frac{1}{p_0} \right)^2 & : \quad p(\mathbf{x}) \le p_0 \\ 0 & : \quad p(\mathbf{x}) > p_0 \end{array} \right\} \quad U_{\text{dyn}}(\mathbf{x}, \mathbf{v}) = \left\{ \begin{array}{cc} \lambda (-\cos\theta)^{\beta} \frac{||\mathbf{v}||}{p(\mathbf{x})} & : \quad \frac{\pi}{2} < \theta \le \pi \\ 0 & : \quad 0 \le \theta \le \frac{\pi}{2} \end{array} \right\}$$



### Dynamic Movement Primitives: Park et al. 2008



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### GPU-Enhanced Variable Attractor Dynamics: Kaldestadt et al. 2014



## GPU-Enhanced Variable Attractor Dynamics: Kaldestadt et al. 2014





Circular Fields: Haddadin et. al. 2011

$$F = \dot{x}_r \times B = \dot{x}_r \times \sum_i I \frac{c_i \times \frac{\dot{x}_r}{\|\dot{x}_r\|}}{l_i^2} da_i$$





Haddadin & al., IFAC2011



### Circular Fields: Haddadin et. al. 2011



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Haddadin & al., IFAC2011





### Cost Based Multi-Agents + Dynamical Systems



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Haddadin & al., Safecomp2013



### How can we ensure that a motion is **safe for a human**?





### Concretely: Safety Analysis of Human Injury & Pain





Is this safe?





### Crash-Testing in Robotics



Haddadin et. al. RSS 2007







Haddadin et. al. RSS 2007





### How to Connect Injury Data with Robot Motion?





## Embedding Injury Knowledge into Control: Haddadin et al. 2010-2012







Embedding Injury Knowledge into Control: Haddadin et al. 2010-2012





 $(mass, velocity, geometry, body part) \rightarrow observed in jury/pain$ 





# Embedding Injury Knowledge into Control: Haddadin et al. 2010-2012





together with Antonio Bicchi, University of Pisa

#### l l Leibniz l O Z Universität l O Ø 4 Hannover



### Example Usecase: Safe Motion Unit



Haddadin et. al. IJRR2012, Haddadin et. al. at 2014, Haddadin et. al. IROS2012, Haddadin & Haddadin Biomecanica Hungaria 2012







Embedding Injury Knowledge into Control: Haddadin et al. 2010-2012

- 1. Assume biomechanical injury/pain data required your application
- 2. Need to know your robot dynamics.
  - Solutions:
    - 1. Ask your robot manufacturer for the data.
    - 2. Do it yourself (e.g. Oussama's PUMA paper, Alessandro's LWR paper).
    - 3. Build your own robot.
- 3. Attach to the safety relevant surface locations of your robot (Points of Interest, POIs) surface primitives that can be associated to your basic injury/pain data
- 4. Run through the set of POIs and calculate their respectively safe velocity.
- 5. Scale the commanded velocity to the most conservative POI.

**Result:** 

### v<sub>max</sub> s.t. safety!!





### Thanks to ... and many others!































### Thanks!

