



A wireless integrated haptic data suit for controlling humanoid robots <u>A. Graziano</u>, P. Tripicchio, C.A. Avizzano, E. Ruffaldi

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**1**.Context and Motivations 2. Proposed Solution **3.**Operator's Body Tracking **4.**Teleoperation and Feedback 5.Evaluation 6.Conclusions

## **Context and Motivations**

- Industry 4.0: Raise of semi-humanoid robots
- ... Cosa ci serve? Cosa manca
- Linea di ricerca
- Online PbD + Cooperation what we need for that?
  - Operator tracking
  - Robot feedback wearable









## Contribution

- We propose an Integrated Suit for:
  - Inertial Based Upper Body Motion Tracking
  - Control a 1-DOF Robotic Gripper with Haptic feedback
- Tested with a teleoperation task of the Baxter (©Rethink) Robot









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## Suit Description

Features:

- Fully portable: batterypowered, Wireless, Modular
- Elastic strips easily wearable for different sized people
- Modularity:
  - O Motion Capture System
  - **O Haptic Gripper Controller**





#### System Architecture Overview

#### For each arm:



#### **Custom Designed Electronics**

- STM ARM based 32 bit  $\mu$ C
- Broadcom WiFi Chip
- Compact: 5cmx5cm
- Versatile Board:
  - Can read IMU via SPI
  - Can Control a DC motor by reading a digital encoder and sending a PWM signal







## **Employed Sensors and Actuators**

Motion Capture System:

Inertial sensors (Invensense 9250):

- 9 sensors
  - 3 accelerometers
  - 3 gyroscope
  - 3 magnetometers
- 16bit resolution on each axis of acc. and gyro.
- 14bit resolutin on each axis of mag.
- Gyro. full scale range:  $\pm 250 \frac{deg}{s}$
- Accel. full scale range:  $\pm 2g$
- Mag. Sensitivity  $0.6\mu T / LSB$

Haptic device:

#### DC motor:

- Highly reduced (300:1)
- 16 count Encoder of motor shaft
- Low voltage power supply (5V)

#### STM integrated H-bridge

• 20kHz max PWM frequency



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## Algorithms: Arm Pose Estimation

- 7DoF kinematic chain to model human arm
- Joint angles obtained by estimating relative attitudes of the sensor frames

Frame	$a_i$	$\alpha_i$	$d_i$	$ heta_i$
1	0	$\frac{\pi}{2}$	0	$\theta_1 + \frac{\pi}{2}$
2	0	$\frac{\pi}{2}$	0	$\theta_2 - \frac{\pi}{2}$
3	l <sub>ua</sub>	0	0	$\theta_3 + \frac{\pi}{2}$
4	0	$\frac{\pi}{2}$	0	$\theta_4 + \frac{\pi}{2}$
5	0	$\frac{\pi}{2}$	$l_{fa}$	$\theta_5 + \frac{\pi}{2}$
6	0	$\frac{\pi}{2}$	0	$\theta_6 + \frac{\pi}{2}$
7	0	0	0	$\theta_7$





## Algorithms: IMU attitude estimation

- The attitude estimation of the sensor frame is performed by using the filter presented in Madgwick et al. [1]
- The filter estimates the attitude by combining:
  - Integration of the angular velocity
  - Minimization of the difference between the expected magnetometer and accelerometer measurements and the actual ones. The minimization is performed with gradient-descent method.



## Algorithms: Compass calibration

- Soft-Iron and Hard-Iron distorsions.
- Static Calibration Vs. Dynamic Calibration
- Calibration can be lost if the user moves from the calibration place.
- An affine model can describe the distortion error:

$${}^{S}\widehat{m} = A_{s} {}^{S}m + {}^{S}c \qquad {}^{s}c = \begin{bmatrix} {}^{s}c_{x} \\ {}^{s}c_{y} \\ {}^{s}c_{z} \end{bmatrix} {}^{s}m = \begin{bmatrix} {}^{s}m_{x} \\ {}^{s}m_{y} \\ {}^{s}m_{z} \end{bmatrix}$$



## Algorithms: Compass calibration

Kalman filter to solve the IMU calibration problem

State and Measurement models:

Magnetometer hard-iron calibration

### Algorithms: Compass calibration

The final estimation equations are:

$$x_{k+1}^{-} = x_{k}^{-}$$

$$x_{k+1}^{+} = x_{k}^{+} + L(0 - Cx_{k}^{-})$$

$$R = \begin{cases} \frac{1}{\|\omega\|}, & \text{if } \|\omega\| < l \\ \frac{1}{l}, & \text{otherwise} \end{cases}$$

$$L = Q_{k}C \left(C^{T}Q_{k}C + R\right)^{-1}$$

 $Q_{k+1} = Q_k - LC^T Q_k$ 

 $Q_k$  is the estimation covariance at step kR is the measurement error covariance

# Algorithms: Sensors orientation calibration



$${}^{A}_{B}\hat{q} = \min_{\substack{A \\ B}q} \sum_{j=1}^{m} \sum_{k=1}^{3} \left\| {}^{A}_{E}q_{k} - {}^{A}_{B}q_{E}^{B}q_{est,j,k} \right\|$$

with  $(A,B) \in \{\{S_T,S_1\},\{S_U,S_3\},\{S_F,S_5\},\{S_H,S_8\}\}$ 

See also [Peppoloni et al. 2013 SISY]

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## **Robot Teloperation**

3DOF control of the Robot Gripper



A closed loop inverse kinematics scheme (CLIK) with EE Robot Jacobian Pseudo-Inverse matrix was adopted.

$$q_{R,k+1} = q_{R,k} + J^+ \left( x_{H,k} - x_{R,k} \right)$$

To obtain a real-time inverse kinematics we perform only three steps of such iterative equation for each new estimated hand pose

#### Algorithms: Haptic Rendering

Classical Virtual Coupling Control Law between robot gripper and operator fingers

$$F = -K_h (X_f - X_g) - B_h (\dot{X}_f - \dot{X}_g)$$





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#### **Evaluation Reconstruction**

#### Validation with the OptiTrack TRIO system





#### **Teleoperation Setup**



Right arm Tracking and Gripping with Oculus

Right arm Tracking – Left arm Gripping



#### Teleoperation



#### **Conclusions and Future Work**

Issues and Future work

Dual-arm ...

Full-dof safe of the Robot End-Effector



## Questions?

