

# Knowledge Engineering in robotics

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BRICS, Rosetta, euRobotics

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# BRICS, Rosetta, euRobotics

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**Rosetta:**

**euRobotics:**

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## BRICS:

- ▶ software engineering for complex robotic systems
- ▶ **how** to do that in (Eclipse, **MDE**) tool support  
= lots of knowledge engineering

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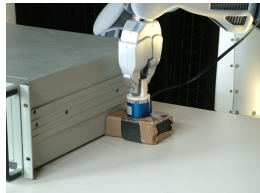
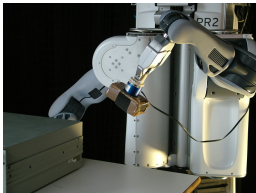
## Rosetta:

- ▶ intelligent skills for force-controlled robotic assembly
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## euRobotics:

- ▶ **semantic web** for robotics portal
- ▶ **need** for *open content* robotics **ontology**

# Examples in my research



# Knowledge representation

**Knowledge** needed is of various **types**:

- ▶ robot motion controllers
- ▶ geometry of objects + “scene graph”
- ▶ sensor capabilities & data interpretation
- ▶ (partial) ordering of actions in task
- ▶ common sense + physical laws
- ▶ relationships robot actions  $\leftrightarrow$  effects
- ▶ ...

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**Representation** of knowledge:

- ▶ (hyper)graphs (Topic Maps, RDF, ...)
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How to **integrate** them...?



# Types of ontologies

- ▶ **object** ontology
  - ▶ what knowledge do our **robots** need to become “intelligent”
- ▶ **domain/system** ontology
- ▶ **profile** ontology

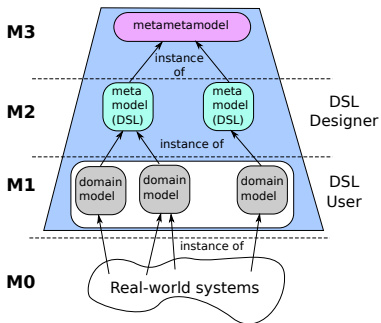
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- ▶ **object** ontology
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  - ▶ what is “*Field robotics*”? Or “*Assembly robotics*”?
  - ▶ Reference: Hallam & Bruyninckx, *An ontology of robotics science*, First European Robotics Symposium, 2006.
- ▶ **profile** ontology

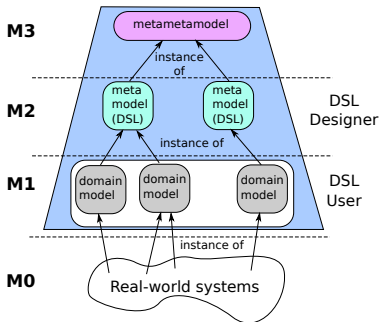
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- ▶ **profile** ontology
  - ▶ what are the competences/expertise of a **researcher**?

# MDE's M0–M3 & ontology

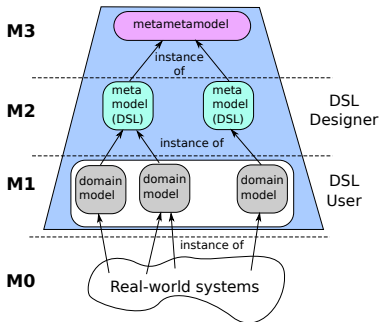


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M0–M3 **is** ontology (not other way around!)

**Claim:** MDE's **Domain Specific Language** concept is **pragmatic way** to start robotics **objects** ontology, in particular for **action representation**



# DSL for assembly case

## —Discrete behaviour: FSM—

```
move_up = apply(tff_motions.move_up, {zt=-0.3}) end
move_down = apply(tff_motions.move_down, {zt=0.1}) end
align = apply(tff_motions.push_down, {zt=10}) end
slide_x = apply(tff_motions.compliant_slide_x,
                {xt=0.2, zt=1}) end
trans:new{ src="initial", tgt="move_down" },
trans:new{ src="move_up", tgt="move_down",
  guard = return get_total_distance() > 0.2 end },
trans:new{ src="align", tgt="slide_x",
  guard = return get_move_duration() > 2 end },
```

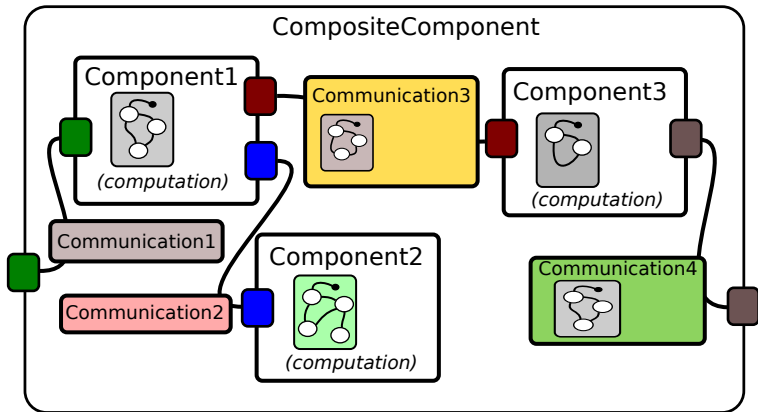
# DSL for assembly case

## —Continuous behaviour: control—

```
move_down =  
  xt = tff.axis_spec:new { value=0 , type='velocity' }  
  yt = tff.axis_spec:new { value=0 , type='velocity' }  
  zt = tff.axis_spec:new { value=0.01 , type='velocity'  
})
```

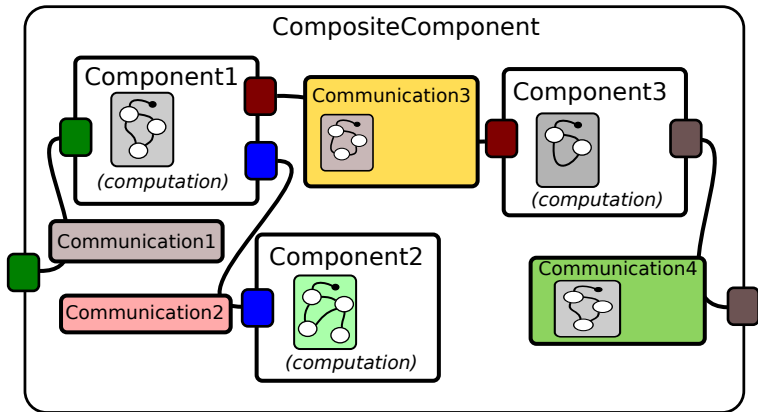
```
compliant_slide =  
  xt = tff.axis_spec:new { value=0, type='force' }  
  yt = tff.axis_spec:new { value=-0.03, type='velocity'  
  
  zt = tff.axis_spec:new { value=1, type='force' }
```

# Robot systems: M2–M3 model



*Structural model + Communication + Coordination*

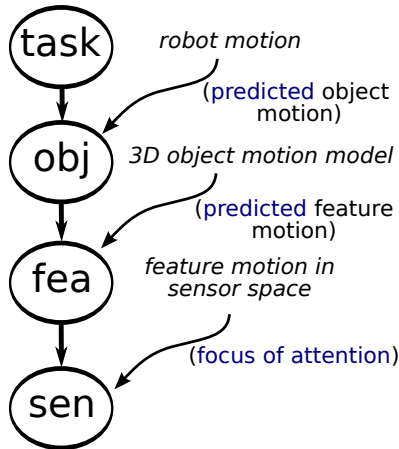
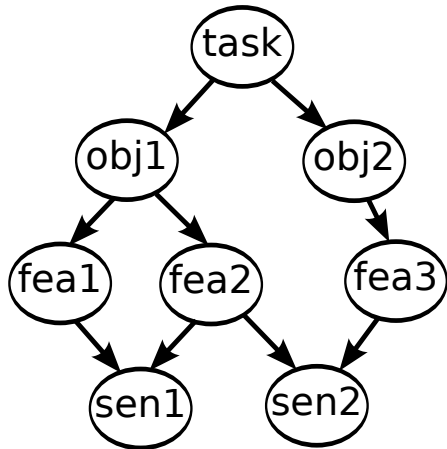
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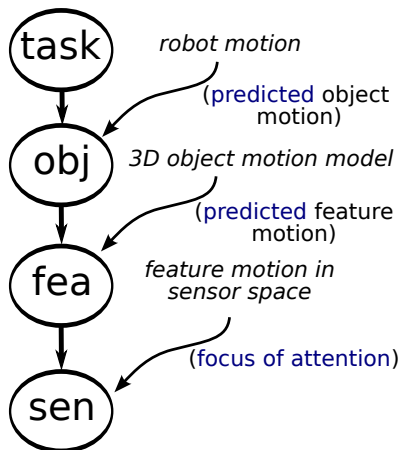
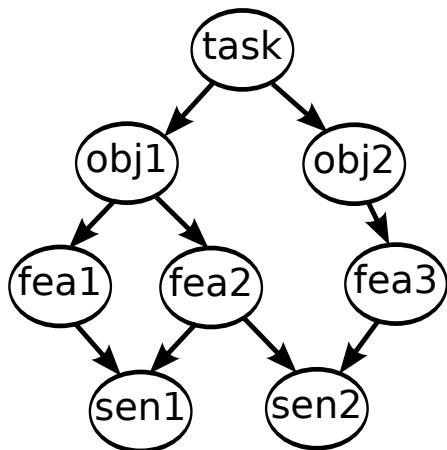
*Structural model + Communication + Coordination*

- ▶ Components: control, learning, planning,...
- ▶ M0–M1 framework DSLs: **Orocos** + **ROS**

# 3D perception stack: M1–M2 model

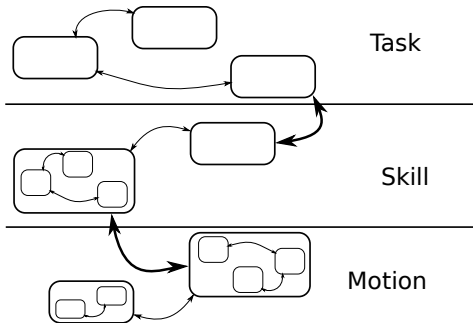


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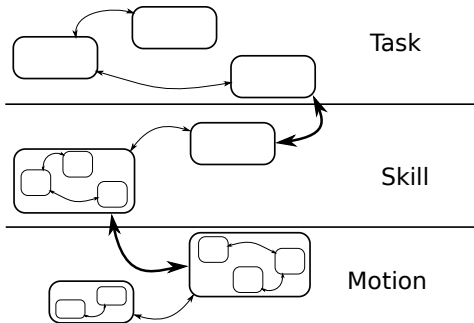


Bayesian probability excellent candidate for DSL!

# Skills: M2–M3 model



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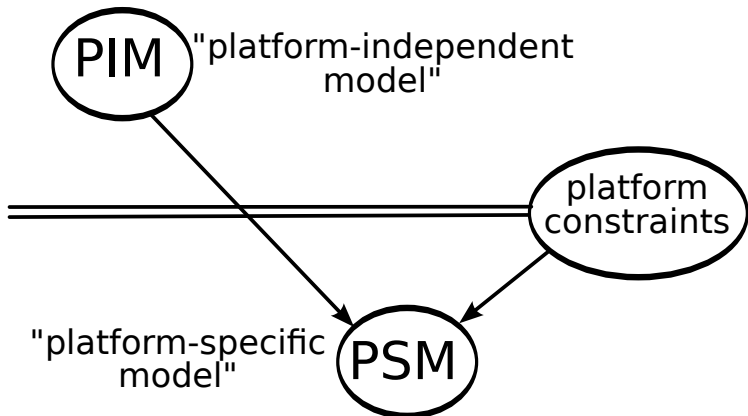


The **Skill** is a **probabilistic state machine**:

- ▶ state machine encodes **causality**/(partial) ordering
- ▶ **events** couple the **symbolic** and **continuous** domains.

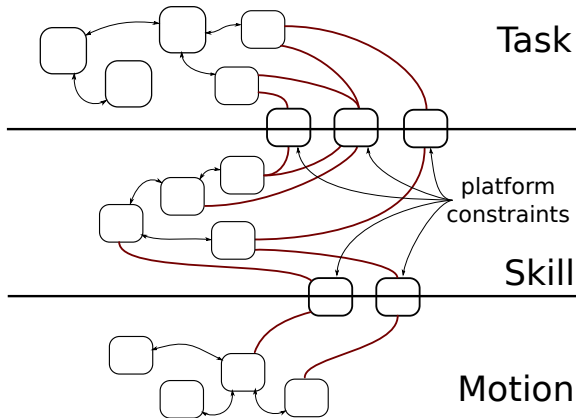


## Skills: M2–M3 model (2)



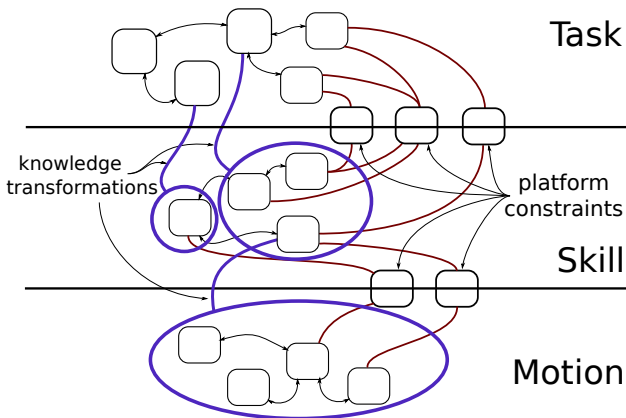
To add **knowledge** on: robot, controller, sensor, learning algorithm,...

# Skills: M2-M3 model (3)



The platform constraints define **parameters** in the FSM **behaviour**.

# Skills: M2-M3 model (4)



The Skill states **are** instantiations of logic symbols, and **run** continuous time/space control & sensing algorithms.

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  - ▶ **focus** about what knowledge and learning to use, at each moment in a robot's task
  - ▶ **grounding & closing the world**: “obvious”
  - ▶ lends itself very well for **DSL** representation
- ▶ Need for **open content** publicly available **ontology server!**

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- ⇒ multi-project cooperation can start **now!**
- ⇒ what **license** shall we use...?  
(Creative Commons–Share alike!?)