Deliberate Action in Robotics: A Perspective

Malik Ghallab



Aug. 2012

Robotics: a driving paradigm for AI

Pioneering AI projects



Outline

3

- Robotics: impressive multidisciplinary achievements
- Robotics: challenges for deliberation functions
 - Motivation for deliberation
 - Spectrum of deliberation functions
- Planning
- Acting
- Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Robotics: a wide multidisciplinary field

Impressive developments of

- Sensory motor functions
- Sophisticated platforms
- Achievements in particular in
 - Dynamic control
 - Motion planning and control
 - Simultaneous localization and mapping
 - Learning sensory motor functions by Reinforcement and/or Demonstration

Dynamic Control



[DLR, Munich]

Dynamic Control Principles: $X_t, U_t \longrightarrow \hat{X}_{t+1}$ Prediction Optimization $\hat{X}_{t+1}, X_{t+1} \longrightarrow U_{t+1}$ Robot's hardware Perceived environment dynamics Exter. sensors Desired Command behavior Inverse Planning Actuators Environment model Effector -> Joints Priop. sensors Perceived Joints -> Effector Forward model Predicted Discrepancy

6

Dynamic Motion Planning



Dynamic Motion Planning L Inverse Legs 1 $X_{ads} X_{hds}$ Feasibility **Feet Trajectory Kinematics** $O_w O_h$ R₁ Unit (FU) (IK) Generator (FTG) W_1 S_w S_h ZMP_{des} $\mathsf{Z}_{\mathsf{hds}}$ **Upper Body** W_1 UB, Motion **CoM and Waist** Generator Traj. Generator (UBMG) CoM₁ (CWTG 1) ZMP_{des} L W_2 **CoM and Waist Feet Trajectory** R Traj. Generator Adaptator (FTA) CoM₂ (CWTG 2) W L_1 [Perrin & Stases, LAAS, Toulouse] 8

Reinforcement Learning of sensory motor functions



Helicopter Aerobatics Apprenticeship Learning



Aerobatics Apprenticeship Learning

Simple linear rigid dynamic models of helicopter

- Learn dynamic models, one for each type of maneuver
 - Regression from teacher's demonstrations
 - Improvement by reinforcement learning in autonomous flight
- ▶ Learn reference trajectories, one for each aerobatic figure
 - Expectation-Maximization on teacher's demonstrations
 - Temporal alignment and optimization
- Learn controllers, one for each aerobatic figure
 - *Differential dynamic programming* continuous MDPs solved by iterative approximation of receding horizon LQR problems

Outline

11

✓ Robotics: impressive multidisciplinary achievements

- Robotics: challenges for deliberation functions
 - Motivation for deliberation
 - Spectrum of deliberation functions
- Planning
- Acting
- Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Robotics: challenges for deliberation functions

- Characterization of a robotics demo
 - Type and variability of tasks and environments
 - Degree of autonomy

Deliberation: not needed if there is

- No variability
- No autonomy





[Curiosity, Mars Space Lab, NASA/JPL]

- Complex mission: geology, atmosphere, climate, environment and bio-environment sciences
- Sophisticated instruments: micro, spectro & telescope, diffraction & radiation detection, soil sampling and analysis
- Extended motion and long range navigation

Motivations

Mission

- Navigate, map, explore, sample, analyze
- Communicate, control instruments and resources
- Mission specification
 - Set of objectives, constraints and choice criteria
 - Not a set of executable commands



Mission

Motivations

- Household services: maintain, sort, clean, cook, etc.
- Support users, help, remind, monitor, etc.
- Interaction
 - Complex multimodal dialogue, plan and intention recognition
 - Cooperation in task achievement



[LAAS, Toulouse]



Deliberate action

17

- Purposeful action
 - Planned, intended to achieve some objectives
 - Pursued for the accomplishment of the robot's mission
- Computational gap between planning and control models

=> Hierarchy of deliberation functions

Integrate action to sensory-motor platform and environment

=> Diversity of deliberation functions



Outline

✓ Robotics: impressive multidisciplinary achievements

 $\checkmark {\sf Robotics:}$ challenges for deliberation functions

- Motivation for deliberation
- Spectrum of deliberation functions
- Planning
- Acting
- Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Planning for deliberate action

- Significant developments in automated planning
 - Several orders of magnitude in classical planning performances
 - Numerous extensions in representations & reasoning capabilities
- Research challenges
 - Integration of planning to acting and interacting
 - Integration of planning to sensing and observing
 - Concurrency of actions and exogenous events

Planning with timelines on state variables



Multi-Robots Planning

Distributed planning, coordination by *plan merging* over shared resources



[LAAS-CNRS]

Multi-Robots Planning





Multi-Robots Manipulation Planning





✓ Robotics: impressive multidisciplinary achievements

- $\checkmark \mathsf{Robotics:}$ challenges for deliberation functions
 - Motivation for deliberation
 - Spectrum of deliberation functions
- ✓Planning
- Acting
- Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Integration of planning and acting





Planning and acting









Selection and switching between skills achieving an action



Navigation



[LAAS, Toulouse]

33

Planning and Acting: approaches

- Temporal approaches IxTeT-eXeC, IDEA, T-ReX
 - CSP based
 - Dispatching, plan extension, revision
- Imperative approaches PRS, RAP, TDL, XFRM, etc.
 - Flexible
 - Hand-written
- State machine approaches PLEXIL, Smach, etc.
 - More open to V&V
 - Hand-written
- Stochastic approaches (MDP, DBN/DDN,)
 - Learned from experiences or from teaching





Outline

✓ Robotics: impressive multidisciplinary achievements

 $\checkmark {\sf Robotics:}$ challenges for deliberation functions

- Motivation for deliberation
- Spectrum of deliberation functions
- ✓Planning
- ✓Acting
- Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Monitoring

35

Role

- Detect predictions observations discrepancies
- Explain and recognize
- Repair
- Approaches
 - Hand-written skills to monitor applicability/maintenance conditions
 - Monitor plan invariants
 - Synthesize invariants from extended planning problems which guarantees the plan execution
 - Model checking execution traces with LTL
 - RMPL: Constraint Based Automata + Control programs
 - TALPlan: synthesize monitors for its plan

Temporal Action Logic for planning and monitoring

$$\begin{split} & [t,t'] \textbf{fly-to}(uav, newx, newy) \rightsquigarrow \\ & [t] \textbf{fuel}(uav) > \textbf{fuel-usage}(uav, \textbf{x}(uav), \textbf{y}(uav), newx, newy) \rightarrow \\ & R([t+1] \ \textbf{hovering}(uav) \triangleq False) \land \\ & R((t,t'] \ \textbf{x}(uav) \triangleq newx \land \textbf{y}(uav) \triangleq newy \land \\ & \quad \textbf{fuel}(uav) \triangleq value(t, \textbf{fuel}(uav) - \\ & \quad \textbf{fuel-usage}(uav, \textbf{x}(uav), \textbf{y}(uav), newx, newy))) \land \\ & t' - t = value(t, \textbf{flight-time}(uav, \textbf{x}(uav), \textbf{y}(uav), newx, newy)) \end{split}$$

TALPIan: order-sorted temporal logic for state-space planning

37

Monitoring integrated to TALPIan

- Predictions synthesized from *monitor formula* and planning knowledge while planning
 - At the action level
 - At the plan level : causal relations between actions

 $\Box \forall uav.(\mathsf{power}(uav) > M \rightarrow$

 $\mathsf{power} < f \cdot M \bigcup_{[0,\tau]} \Box_{[0,\tau']} \mathsf{power}(uav) \le M)$

- Surveillance of states and sequences wrt monitoring prediction with incremental formula progression algorithm
- Diagnosis of discrepancies
- Error recovery

Outline

✓ Robotics: impressive multidisciplinary achievements

 $\checkmark {\sf Robotics:}$ challenges for deliberation functions

- Motivation for deliberation
- Spectrum of deliberation functions
- ✓ Planning
- √Acting
- ✓Monitoring
- Observing
- Goal Reasoning
- Architecture
- Conclusion

Observing

39

Role

- Process signals needed in closed loop servoing
- Detect and structure environment features, recognize, categorize, link signals to symbols (anchoring)
- Predict from sequences of events ongoing situations, plans and temporal chronicles
- Bottom-up to from signal to symbols
- Top-down to focus attention and trigger observation actions

Observing Anchoring problem • Relate perceptual data and symbolic attribute corresponding to the same physical object $\alpha =$ $, \hat{\gamma} >$ <. individual cup22 symbols percepts predicate {red, large} symbols attributes [-20,20] [220.260] Σ Π [Örebro Univ.] • Track anchors overtime and refine/revise hypothesis [Linköping Univ.] 41

Perception Engine

- Comprehensive and coherent approach for observing
- Data flow architecture
- Stream based formalism on perception processes:
 - primitive, refinement, configuration, mediation processes
 - policies over processes, temporal constraints
- Natural integration to planning and monitoring,
- Opens V&V perspectives



Deliberation functions



<section-header><list-item><list-item><list-item><list-item>
Outline
Robotics: impressive multidisciplinary achievements
Robotics: challenges for deliberation functions
Motivation for deliberation functions
Spectrum of deliberation functions
Planning
Acting
Monitoring
Goal Reasoning
Architecture
Conclusion

43

Goal Reasoning

45

- ► Goal Driven Autonomy (GDA), overall mission perspectives
- Higher level monitoring wrt objectives, criteria and constraints
- Discrepancy detector
- Explanation generator
- Goal generator to address new conditions
- Goal manager
 - Decision theory: tradeoff between conflicting goalls
 - Explicit choice
- Found in large systems (CPEF, DS1)
 Function often embedded in acting/monitoring/planning















Architectures

Outline

 $\checkmark \mathsf{Robotics: impressive multidisciplinary achievements}$

 $\checkmark {\sf Robotics:}$ challenges for deliberation functions

- Motivation for deliberation
- Spectrum of deliberation functions

√Planning

✓Acting

✓Monitoring

√Observing

✓ Goal Reasoning

- ✓Architecture
- Conclusion

49

Deliberation in robotics: key integrative challenge for Al

- Planning & Acting
- Observing the environment semantics
- Monitoring
- Goal reasoning
- Interacting
- Learning
 - Models of the robot and the environment
 - Categories
 - Functions, skills and behaviors
- Architecture
 - Specification
 - Robust adaptation

Acknowledgments

- ► LAAS, Toulouse
 - Felix Ingrand, Rachid Alami, Olivier Stasse
 - Benoit Morisset, Stéphane Cambon, Fabien Gravot, Severin Lemaignan
- Linköping U.: Patrick Doherty and colleagues
- > DLR, Munich: Berthold Bäuml, Gerd Hirzinger
- TU Munich: Michael Beetz and colleagues
- MPI Biological Cybernetics, Tübingen: Jan Peters and colleagues
- Örebro U.: Alessandro Saffiotti
- Stanford U.: Peter Abbeel, Andrew Ng
- ▶ NASA AMES and JPL: David Smith, Vandi Verma, Jeremy Frank
- MBARI: Kanna Rajan