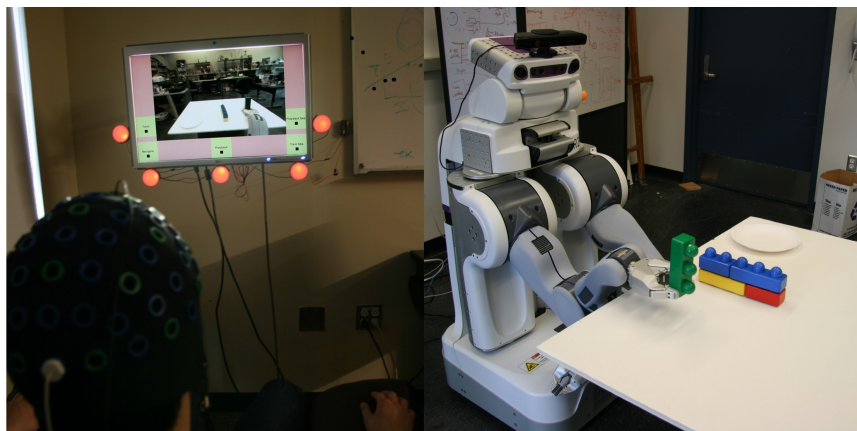


Interactive Hierarchical Brain-Computer Interfacing: Uncertainty- Based Interaction between Human and Robots



5th International BCI Conference 2011

Graz, Austria, September 22, 2011



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Outline

- Hierarchical BCIs
- Uncertainty-based interactive hierarchical BCIs

Traditional BCIs for Robotic Control

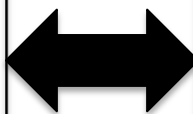
- Trade-off between cognitive load and scalability

High-level control paradigm: more robotic autonomy

low cognitive load

but

coarse-grained control



Low-level control paradigm:
Finer-grained moment-by-moment control

- **High-flexibility**

but

- **higher-cognitive load**

Hierarchical BCIs Phase I: Train

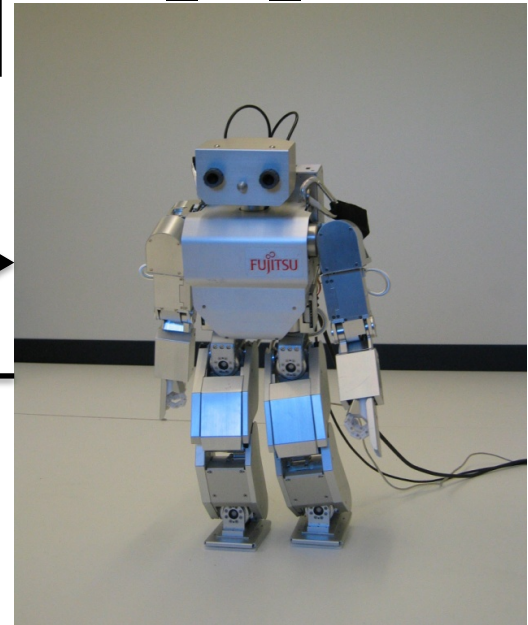
*User intention:
"Go to kitchen"*



*User notices robot
reached goal*

EEG command:
"Learn GO_TO_KITCHEN
skill"

*Available high-level
control commands:
- GO_TO_KITCHEN*



*Robot "learns" a
position-based
navigation skill*

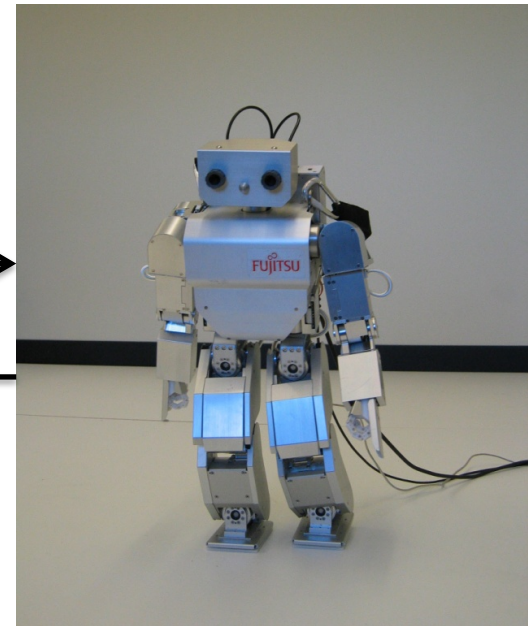
Hierarchical BCIs Phase II: Test

*User intention:
"Go to kitchen"*



EEG command:
"GO_TO_KITCHEN"

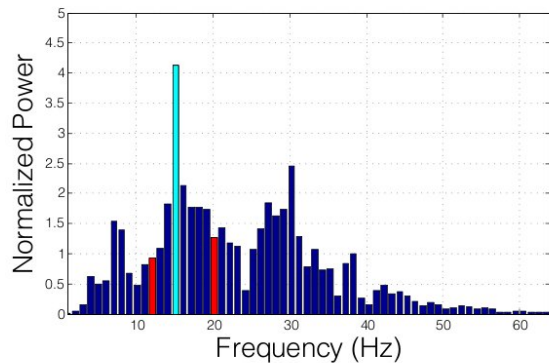
*Available commands:
"GO_TO_KITCHEN"*



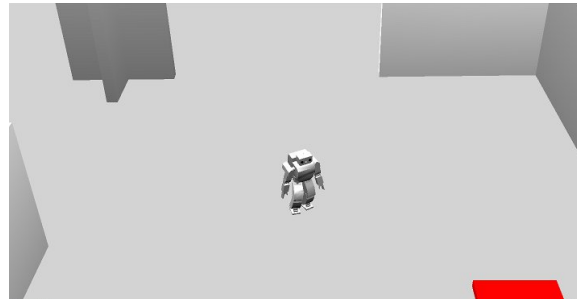
*Robot goes to
kitchen*

System Components

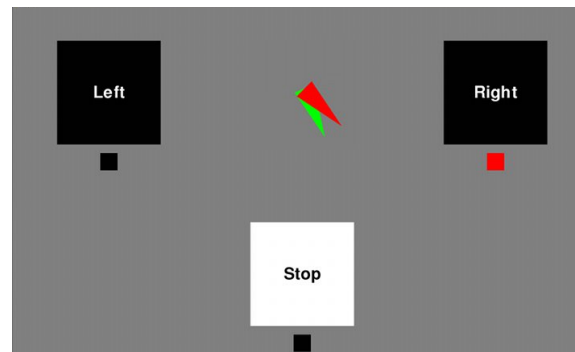
SSVEP-based BCI



Robot Application



Hierarchical Adaptive Menu



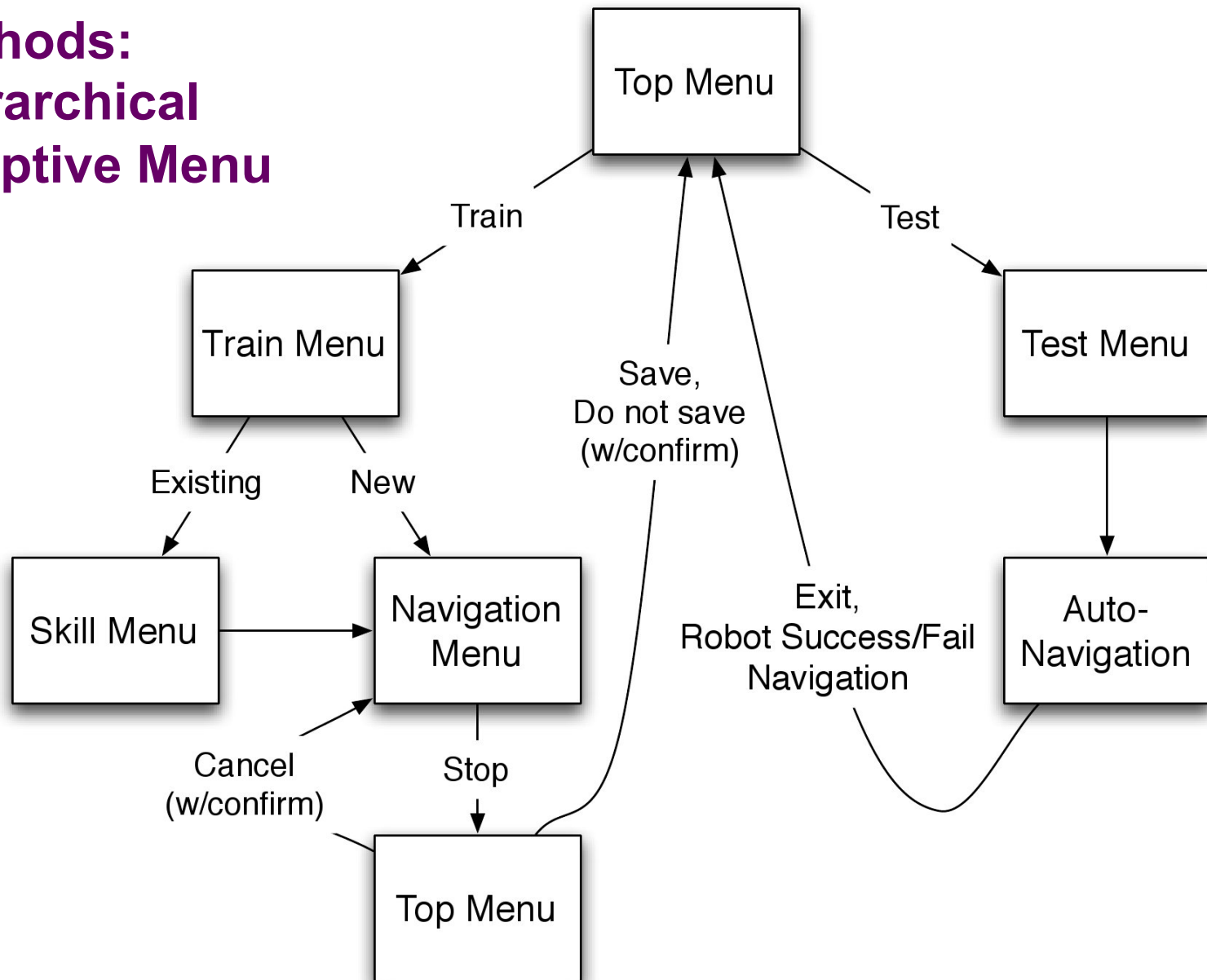
Methods: BCI

- Current system uses SSVEP (but not limited to)
 - TFT monitor with refresh rate of 60Hz.
 - Three options: 12 Hz, 15 Hz, and 20 Hz
 - Asynchronous BCI paradigm (e.g., motor imagery) could be a more natural interface
- Classification
 - data collection (4s), refractory periods (2s)
 - classification using frequency domain features

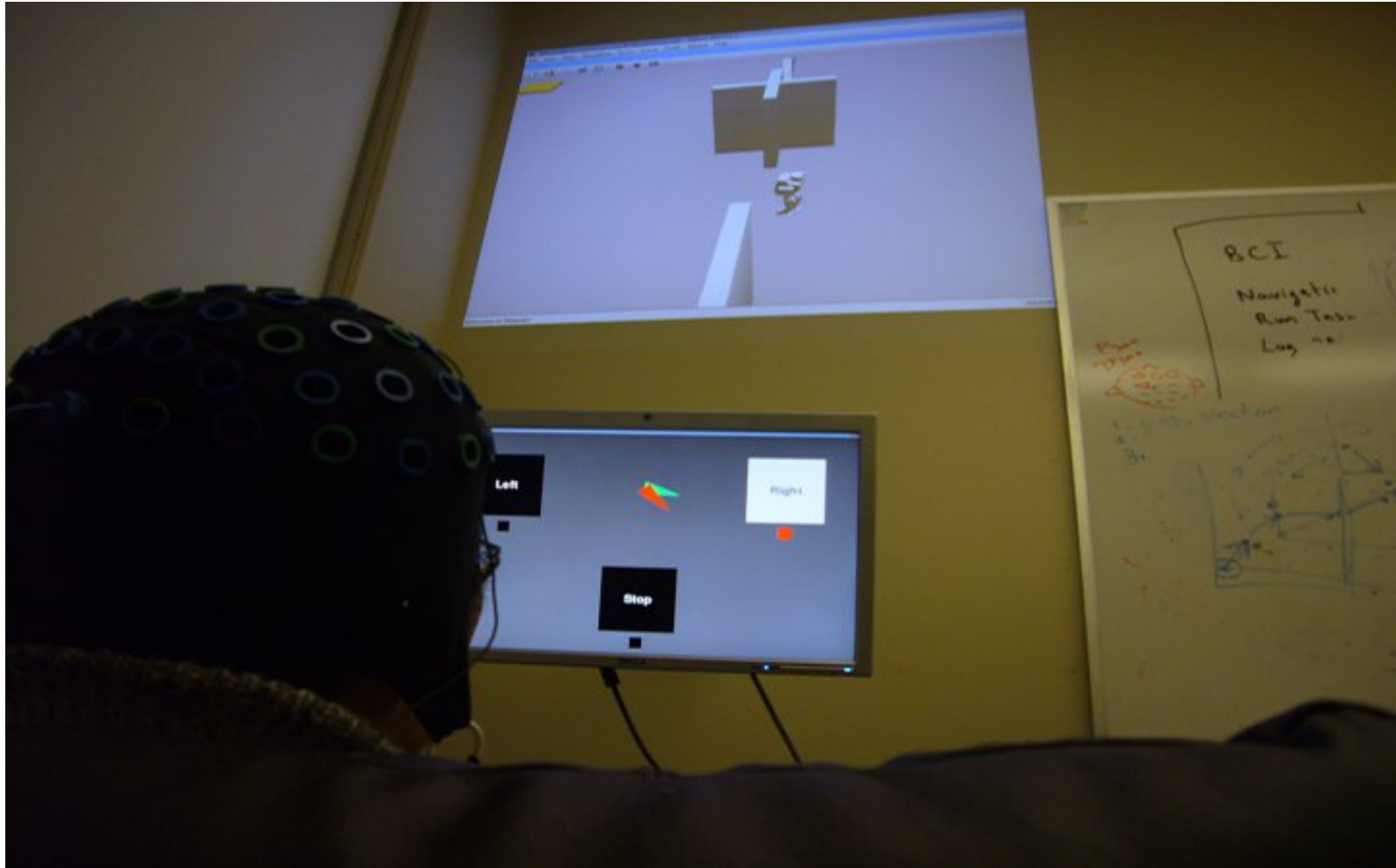
Methods: Robot Learning

- Learning “high-level” control commands on-the-fly from “low-level” control demonstration traces.
 - function approximator, e.g., RBF Neural Network, or Gaussian Process Regression
 - training data: position based traces from “low-level” control demonstrations
 - output of function approximator produces sequence of control commands until goal-state is reached
 - one function approximator for each high-level control command

Methods: Hierarchical Adaptive Menu

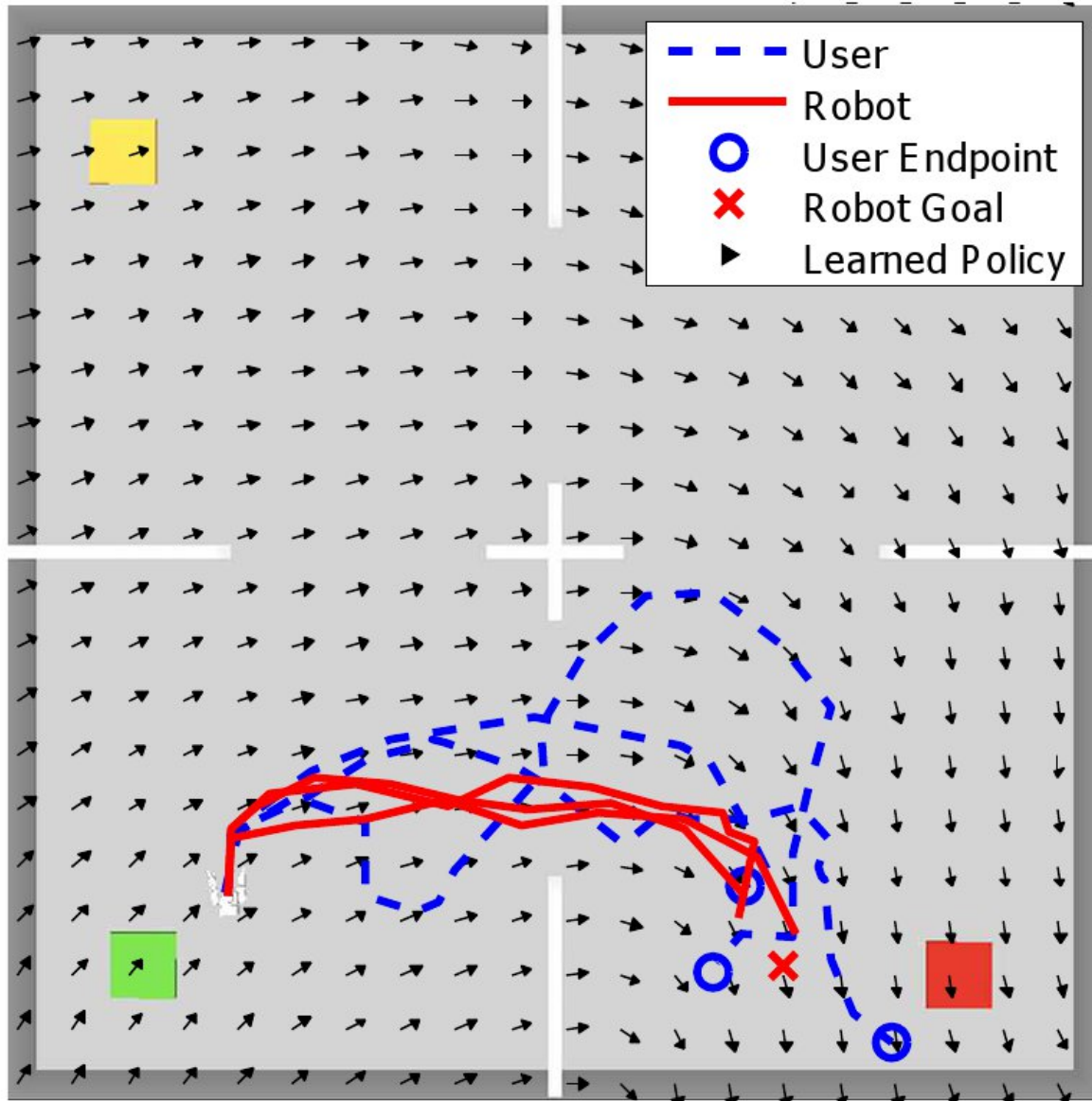


Experimental Setup



Results

Navigation traces and policy



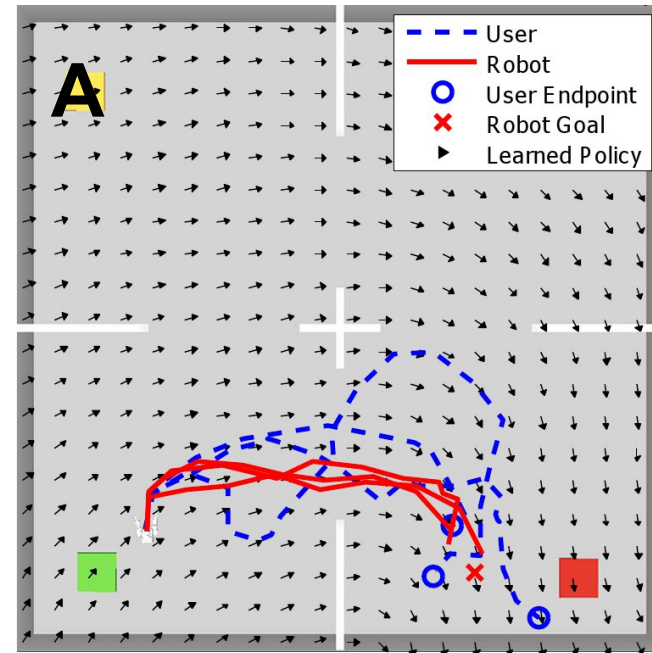
Results

| | Low-level BCI | Hierarchical BCI |
|--|---------------|------------------|
| Mean among four subjects (std) | | |
| Number of selections made | 20 (7) | 5 (2) |
| Task completion time (s) | 220 (67) | 112 (25) |
| Navigation only time (s) | 124 (37) | 73 (19) |
| Mean of three trials from best subject (std) | | |
| Number of selections made | 15 (5) | 4 (1) |
| Task completion time (s) | 141 (42) | 85 (4) |
| Navigation only time (s) | 99 (30) | 74 (9) |
| Minimum (std) | | |
| Number of selections made | 8 | 4 |
| Task completion time (s) | 91 | 75 |
| Navigation only time (s) | 59 | 58 |

Interactive Hierarchical BCIs

- Unreliable “high-level” skills due to incomplete, or insufficient training data
- Example:

Q: What happens if the robot starts from location “A”?

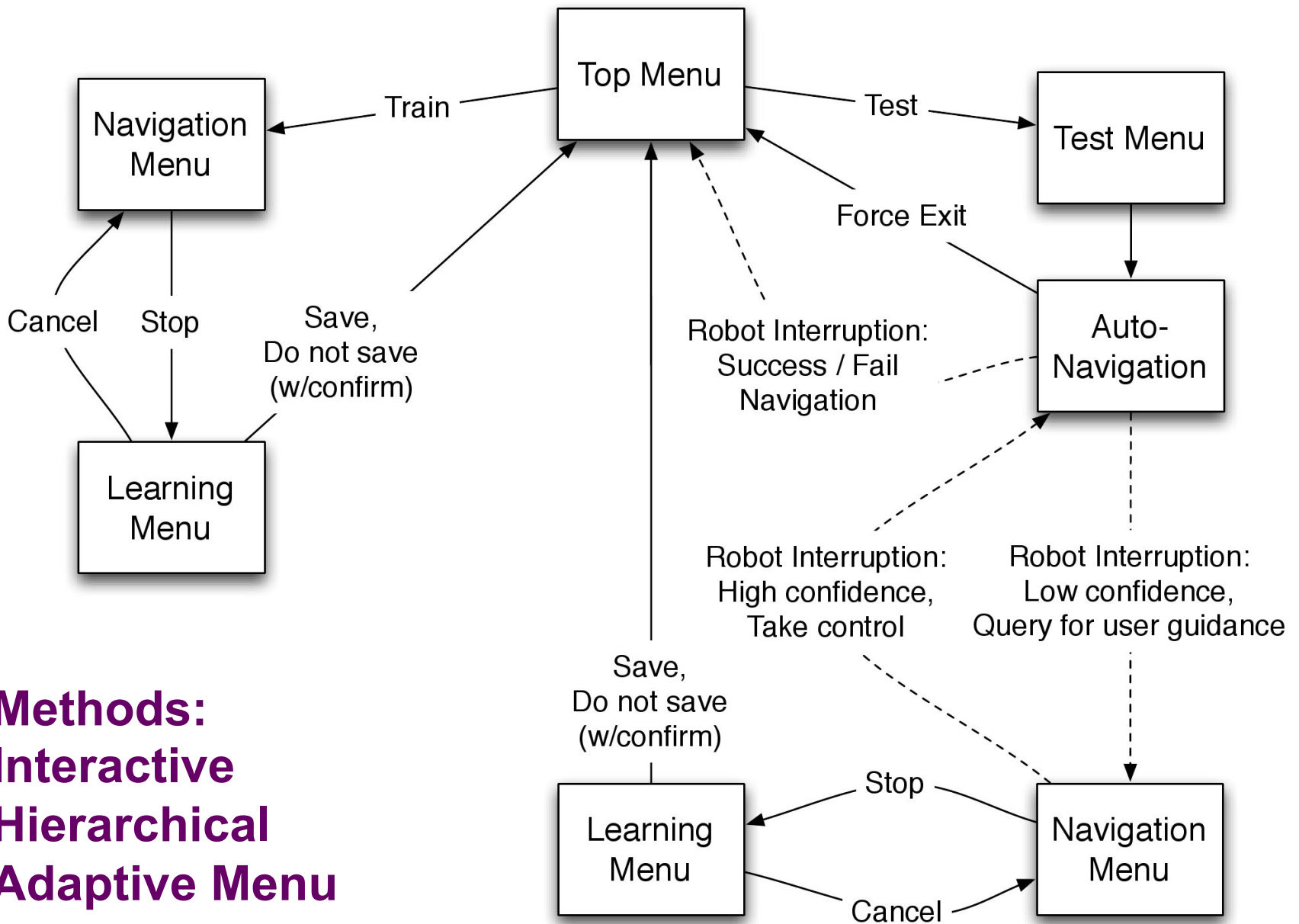


Uncertainty-based Interaction

- High-uncertainty region: Ask for user guidance
 - User gives additional “low-level” control commands to help robot finish the high-level command.
- Low-uncertainty region: Take control from user, autonomously finish an issued high-level command.
 - Relieves the user from engaging in low-level control.

Methods: Robot Learning

- Gaussian Process (GP) function approximator
 - Output of GP: $\langle \text{mean}, \text{variance} \rangle$
 - Variance used as “uncertainty-metric”

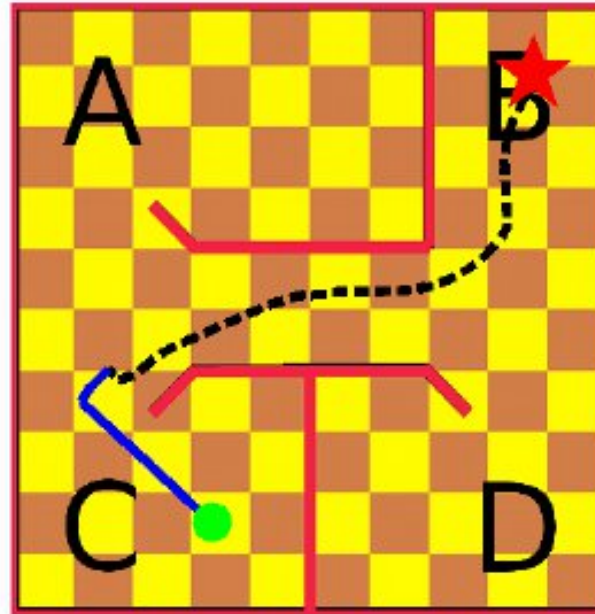


**Methods:
Interactive
Hierarchical
Adaptive Menu**

Result

*Test: High-level command
“Go to location B”*

*Immediately switches
to user demonstration
mode due to high-
uncertainty!
(**blue line trace**)*



*Once the user drives the
robot to low-uncertainty
region, BCI takes control
from user.
(**black dotted line**)*

Red Star: Goal position

Green Dot: Start position

Blue Line: User demonstration trace

Black Dotted Line: Autonomous robot navigation trace

Result

*Updated confidence map after
incorporating more data*



Black Area: Highly uncertain region
White Area: Less uncertain region

Comparison

*Learned confidence map
before update*



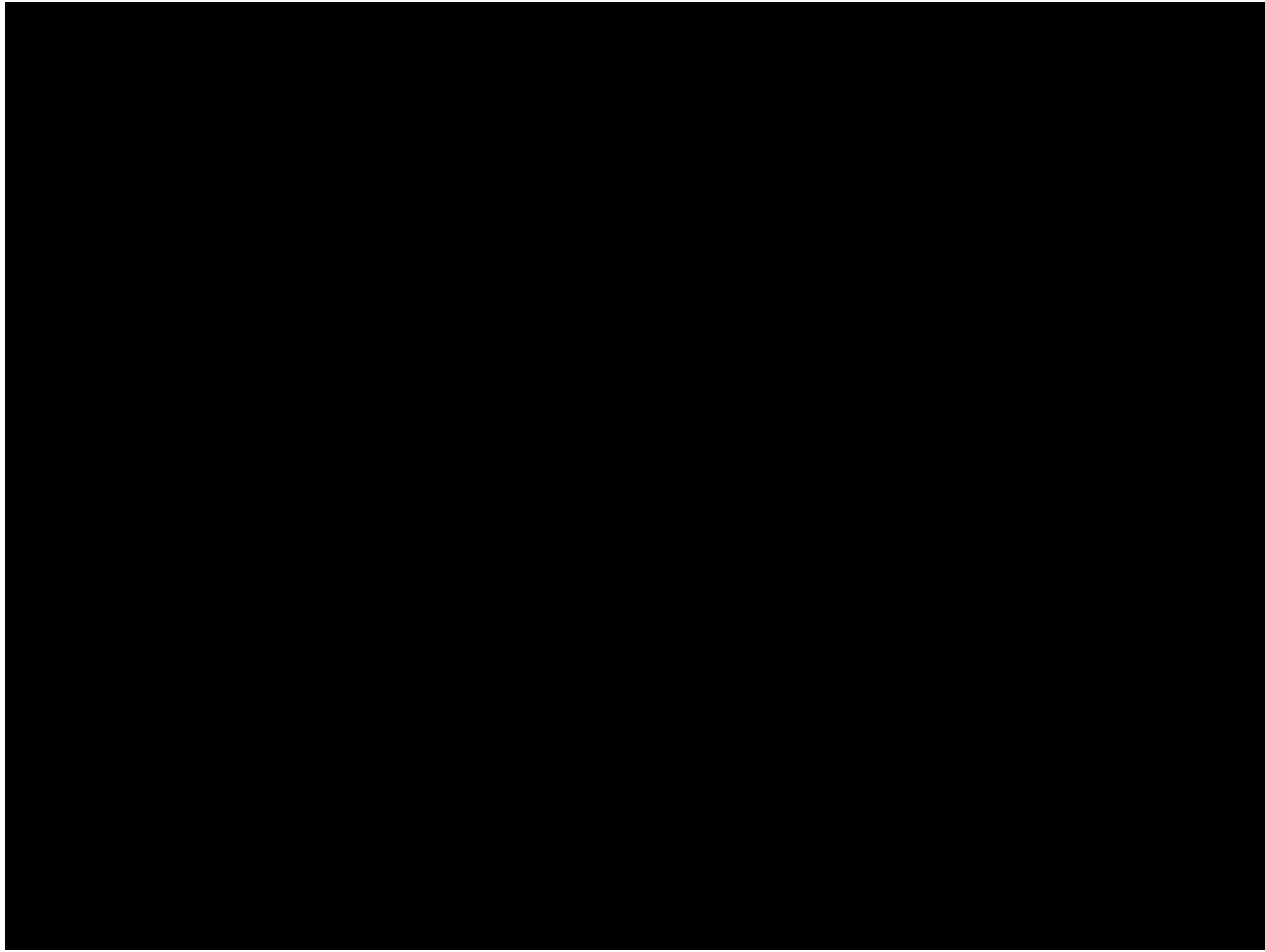
*Updated confidence map after
incorporating more data*



Conclusion

- **Hierarchical BCI**
 - Combines advantages of fine-grained and high-level autonomous control paradigms.
 - Learns high-level commands on-the-fly from user demonstrations with “low-level” control.
- **Uncertainty-based interactive hierarchical BCIs**
 - Interaction based on “uncertainty-metric” makes BCI more reliable and robust while remaining adaptive to user’s needs
 - Ability to handle uncertainty opens the door to practical real-world BCIs

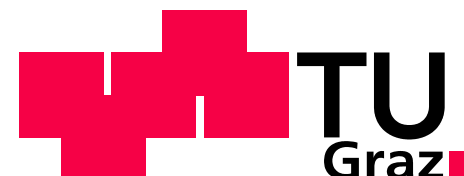
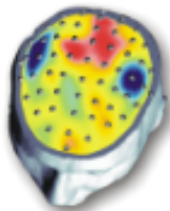
Towards Practical Hierarchical BCIs



Acknowledgments



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Funding: National Science Foundation (0622252 & 0930908), the Office of Naval Research (ONR), and the ICT Collaborative Project BrainAble (247447).

Other: Rawichote Chalodhorn for helping with the HOAP- 2 robot and Webots , Josef Faller for helping with the implementation of the SSVEP stimuli.

