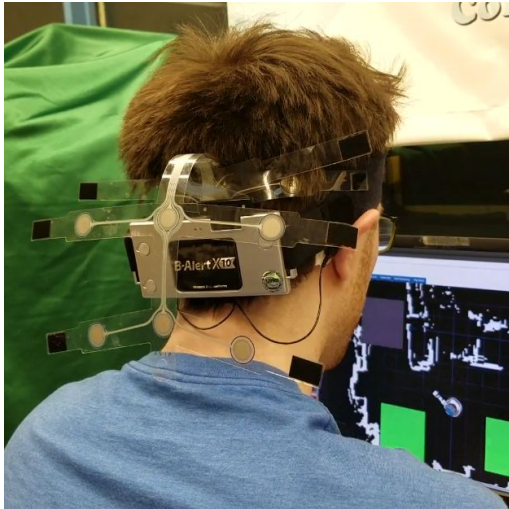


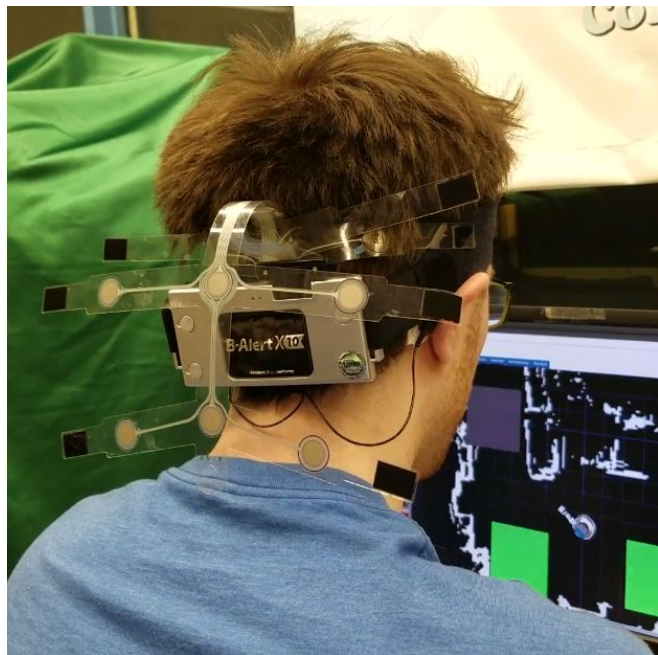
# Task Level Hierarchical System for BCI-enabled Shared Autonomy

Iretiayo Akinola, Boyuan Chen, Jonathan Koss,  
Aalhad Patankar, Jake Varley & Peter Allen  
Columbia University



# Shared Autonomy

## Agent 1



## Agent 2



# Why BCI Interface?

- Robust Assistive Robotics Application
  - Can be used humans with disabilities
- Complementary to other interfaces for complex tasks
  - Expand range of interface modalities
- Move BCI from the lab into real world
  - BCI Robotics Applications e.g.Home-Assistant Robot
  - Spur growth in BCI technologies



# Which BCI?

Different BCI imaging modalities measure brain activity:

- electroencephalography (EEG),
- near-infrared spectroscopy (NIRS),
- magnetoencephalography (MEG),
- functional magnetic resonance imaging (fMRI),
- electrocorticography (ECoG), and
- intracortical electrode recordings

Portable



Non-Invasive



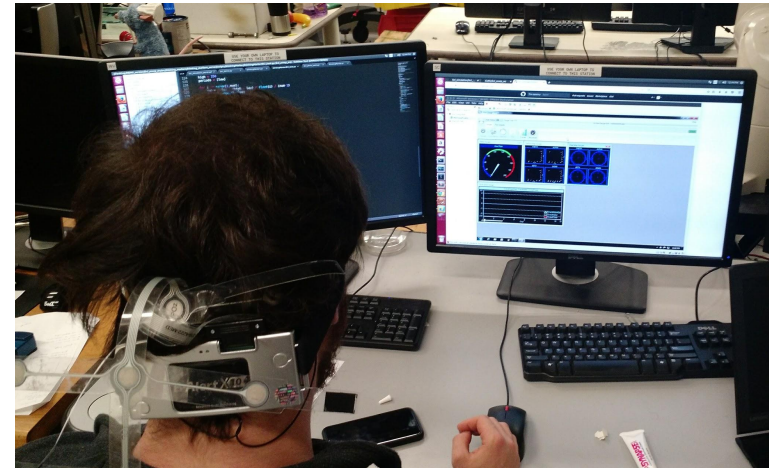
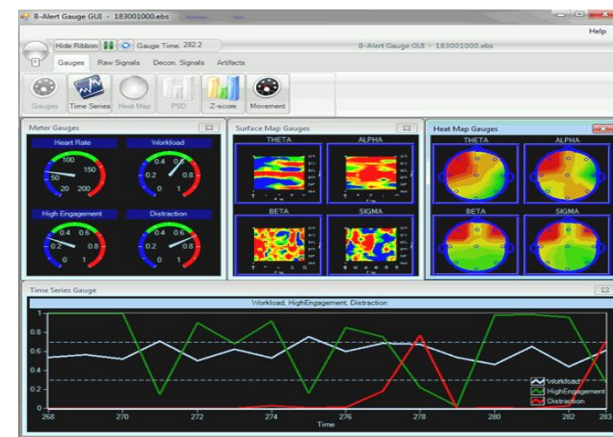
# EEG-BCI Neural Patterns

SSVEP- Steady-State Visual Evoked Potentials

MI- Motor Imagery (Use Sensorimotor Rhythms)

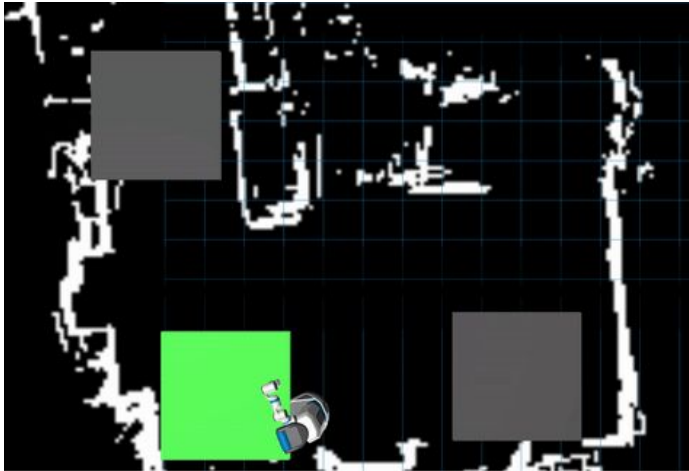
ErrP- Error related Potentials

Affective States



# SSVEP

- Visual Stimulus Driven
- Split frequency band (6-9.5Hz) into # options
- 2 Electrodes in Occipital region (O1 & O2)
- canonical correlation analysis (CCA)



Options are presented to human agent as visual stimuli

# SSVEP Pros & Cons

## Pros

- require no training
- analysis is fairly simple
- reliable and robust response.
- provides high temporal resolution signals for analysis

## Cons

- Requires stimuli
- Discomfort over time
- Small Latencies

# Robot Autonomy

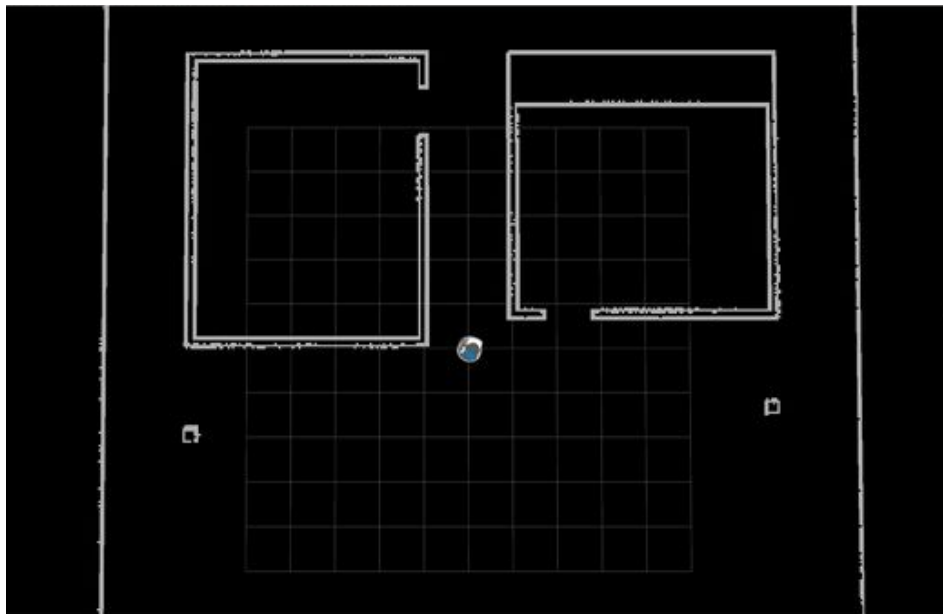
- Navigation
  - SLAM (ROS package)
- Vision Processing
  - Point Cloud Segmentation
  - SSVEP Stimuli generation
- Manipulation
  - Shape completion
  - Grasp planning (Graspl!)
  - Trajectory Planning (MoveIt)





# Robot Autonomy

- **Navigation**
  - **SLAM (ROS package)**
- **Vision Processing**
  - Point Cloud Segmentation
  - SSVEP Stimuli generation
- **Manipulation**
  - Shape completion
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# SLAM- Simultaneous Localization and Mapping

- Mapping- building of a model of the environment
- Localization- estimation of the state of the robot
  - Noisy measurement from sensors (e.g. range sensors, odometry)

## Position state estimation (ACT and SEE cycle)

- SEE: Laser scanner
  - a range of 25m, 220° field of view, 15Hz update rate
  - angular resolution of 1/3°
- ACT: Mobile Base
  - 2 active wheels, 2 free turning wheels
  - Wheel Encoders (resolution not in manual)



# SLAM

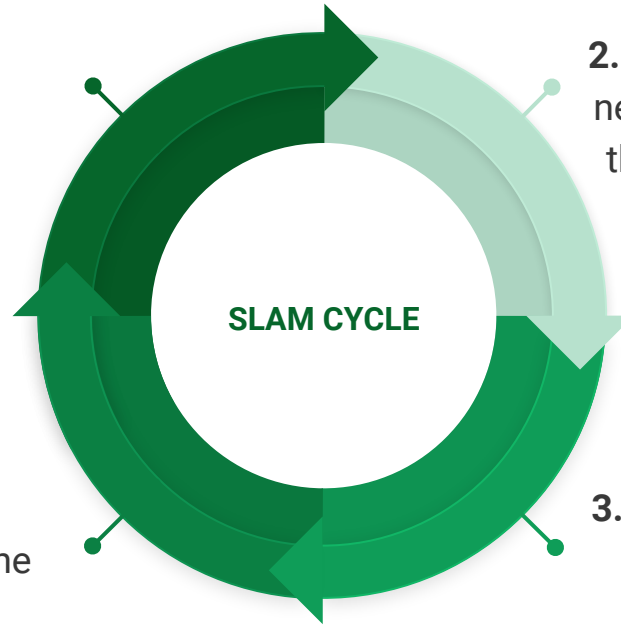
Position state estimation (ACT and SEE cycle)

**1. Sense. Update the estimated state** from registering landmarks. (SEE)

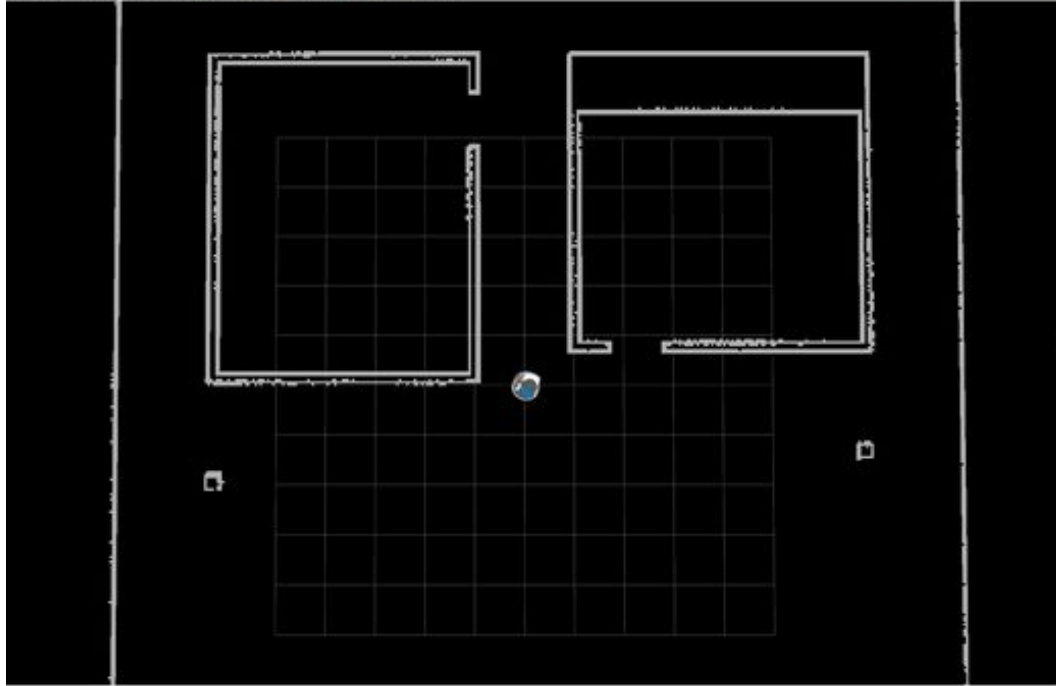
**2. Build Map-** Add new landmarks to the current state.

**3. Move to a new location (ACT)**

**4. Update state estimate** using the odometry data



# Demo: SLAM in Action

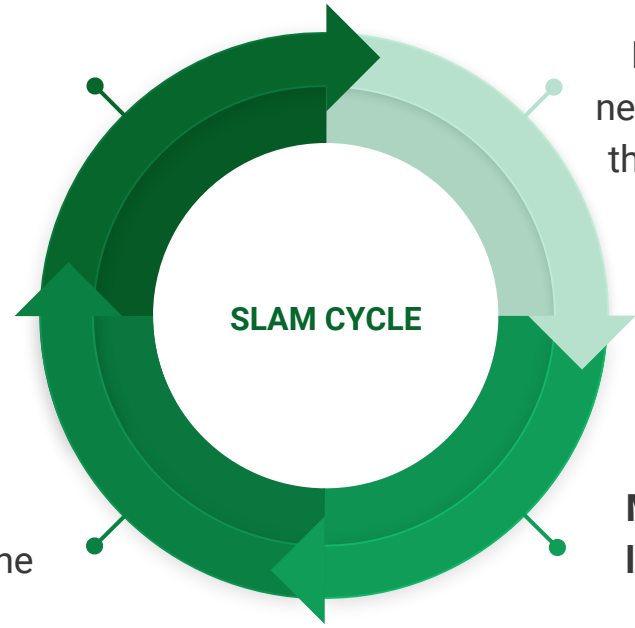


# SLAM

The robot estimates the belief state about its position through an ACT and SEE cycle

**Sense. Update the estimated state** from registering landmarks. (SEE)

**Build Map-** Add new landmarks to the current state.

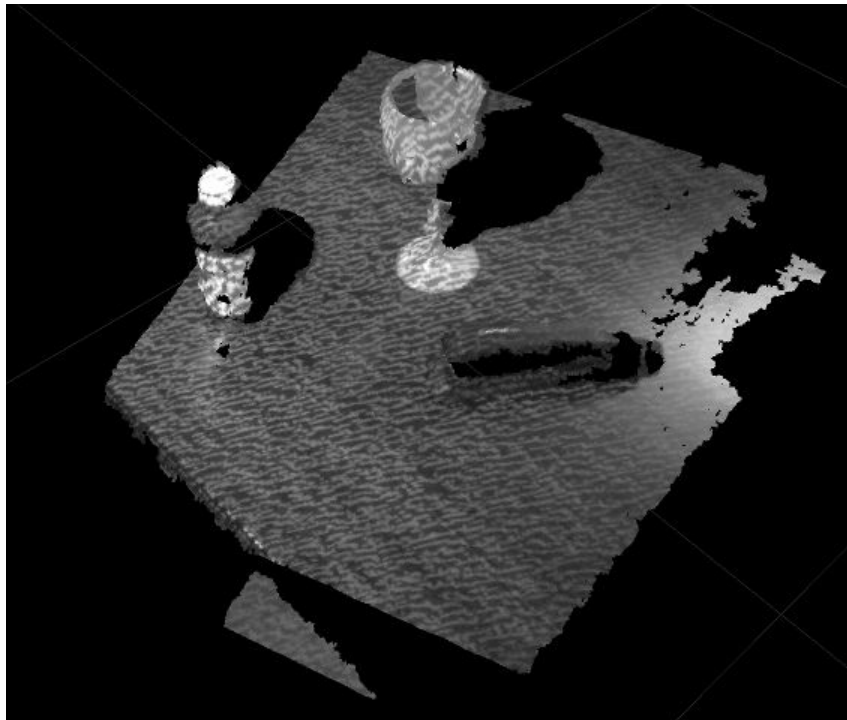


**Update state estimate** using the odometry data

**Move to a new location (ACT)**

# Robot Autonomy

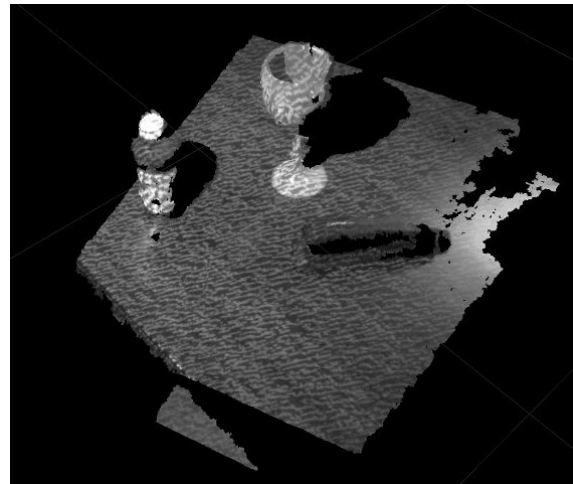
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# Point Cloud Segmentation

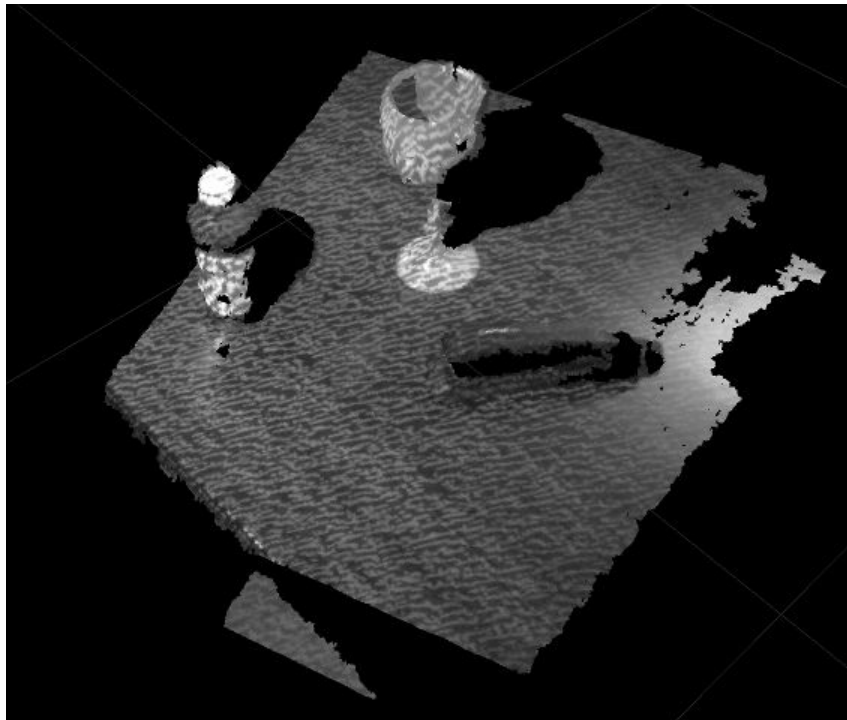
## Euclidean Cluster Extraction

- create a **Kd-tree** representation for the input point cloud dataset  $P$ ;
- set up an empty list of clusters  $C$ , and a queue of the points that need to be checked  $Q$ ;
- for every point  $p_i$  in  $P$ 
  - add  $p_i$  to the current queue  $Q$ ;
  - for every point  $p_i$  in  $Q$ 
    - search for the set  $P_i^k$  of point neighbors of  $p_i$  in a sphere with radius  $r < d_{th}$ ;
    - for every neighbor  $p_i^k$  in  $P_i^k$ , check if the point has already been processed, and if not add it to  $Q$ ;
    - when the list of all points in  $Q$  has been processed, add  $Q$  to the list of clusters  $C$ , and reset  $Q$  to an empty list



# Robot Autonomy

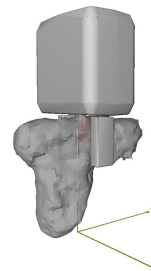
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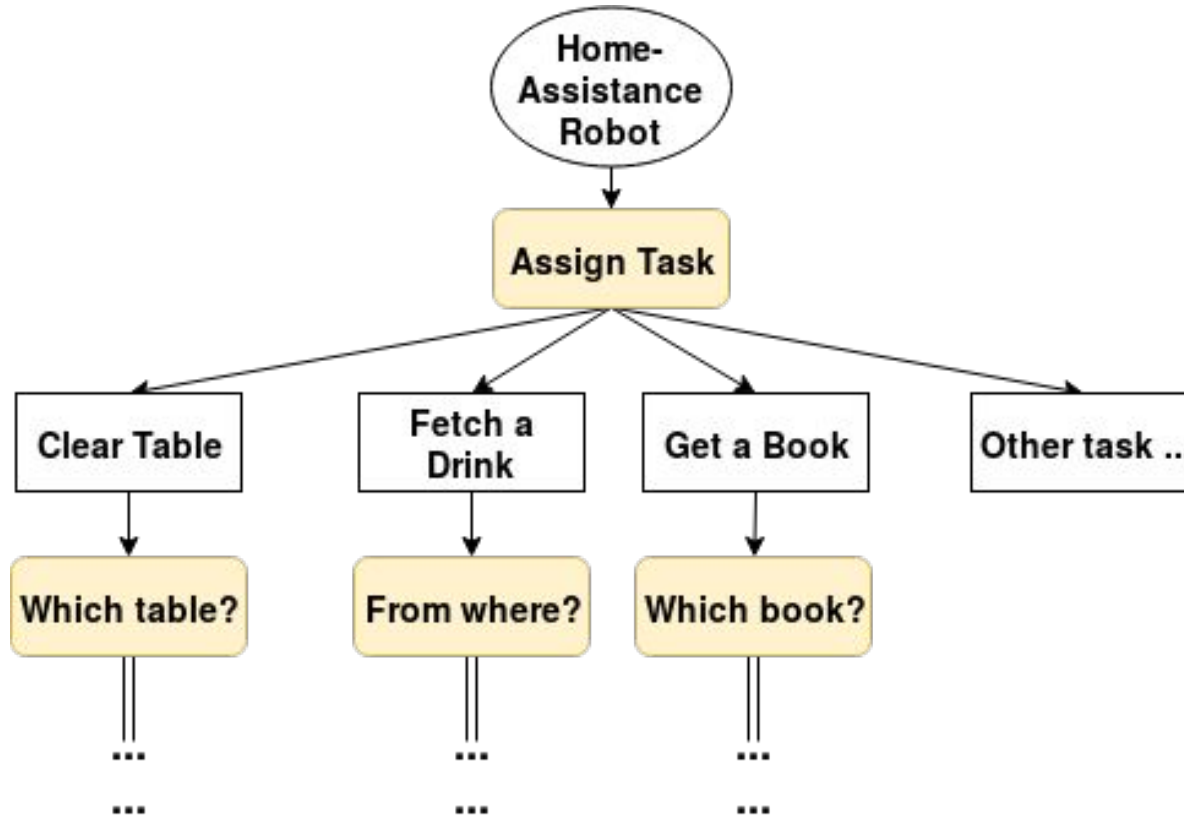


# Robot Autonomy

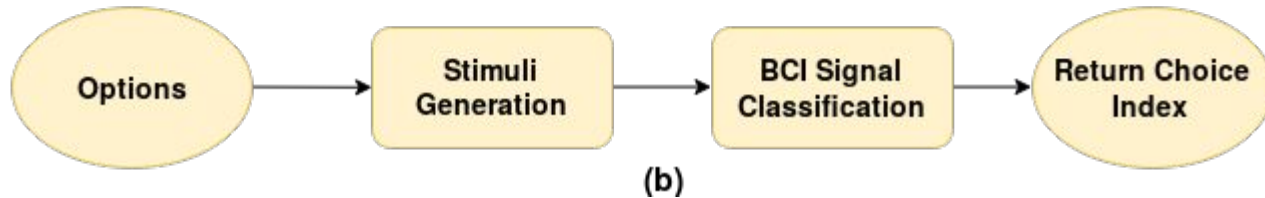
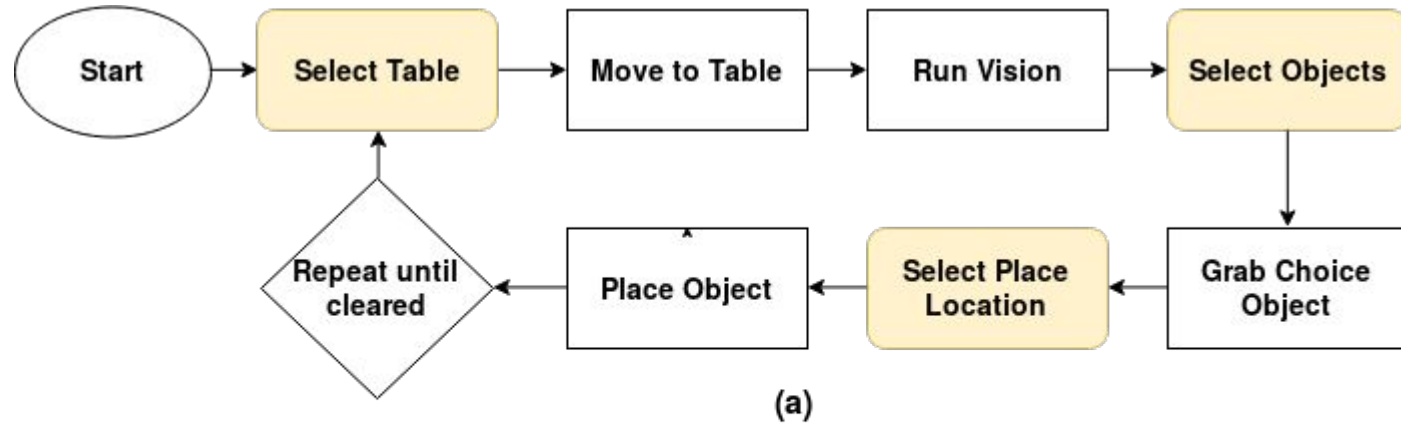
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# Hierarchical System



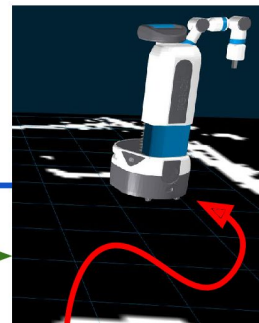
# Hierarchical System- An Instantiation



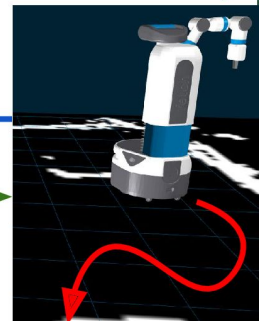
# Table-Clean Up Experiment



Where do I go?  
Go to the table.



What should I grab?  
Pick up the drill.

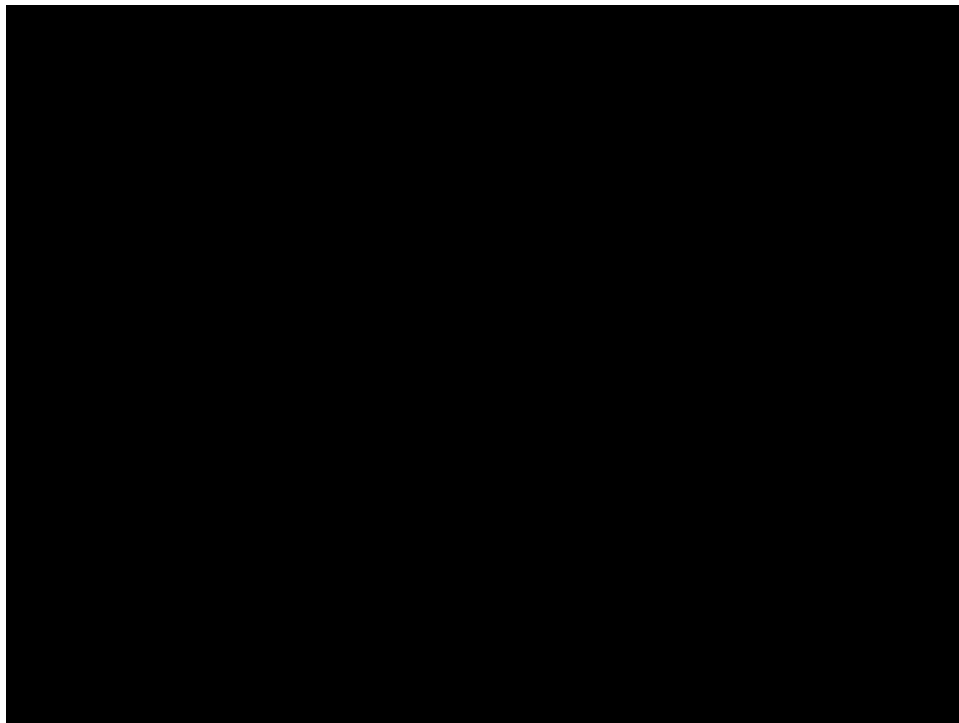


Where do I take it?  
Bring it to me.

...



Video



# Evaluation Criteria

- BCI Success Rate (User Input Detection)
- Mean Time Distribution Between Stages
- Mean Time to Completion
  - ranged from 439s to 543s (mean = 481.3s)

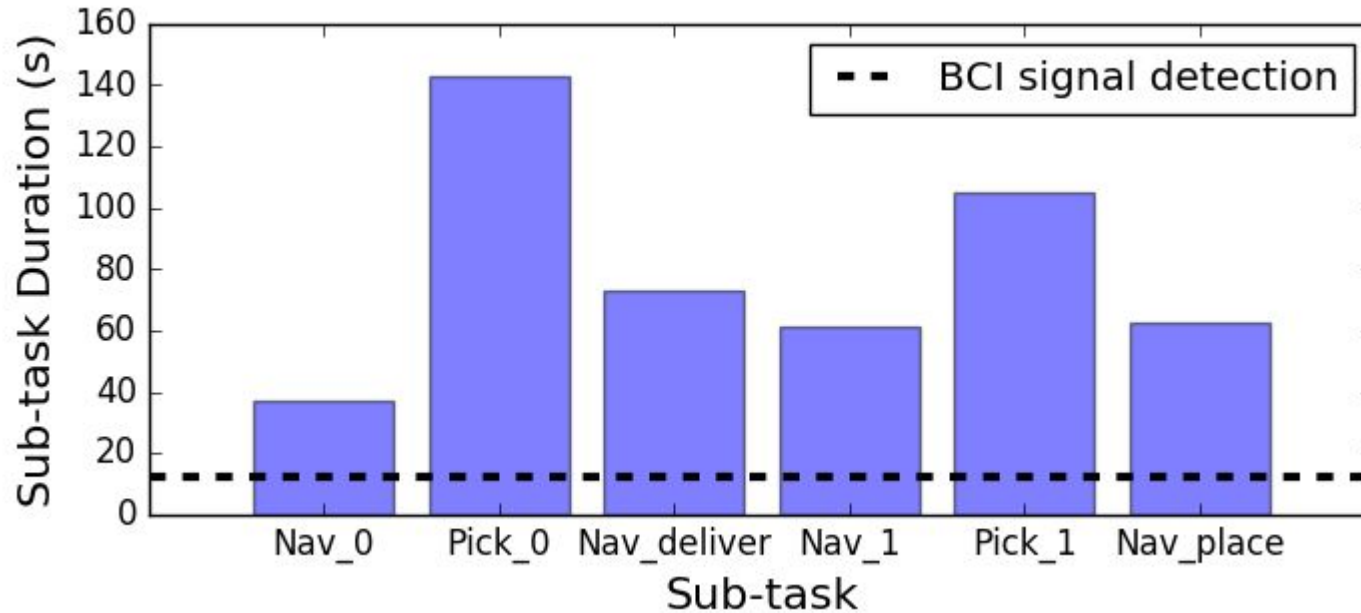
# Results

TABLE II: User study results for table cleanup task.

Subject	# of Trials	SSVEP Classification Success (# successful queries / # queries)
1	3	15/15 (100%)
2	3	11/12 (91.7%)
3	3	11/12 (91.7%)
4	3	14/14 (100%)
5	3	15/15 (100%)
6	3	14/15 (93.3%)
7	3	15/15 (100%)
		Total: 95 / 98 (96.9%)

# Mean Time Distribution Between Stages

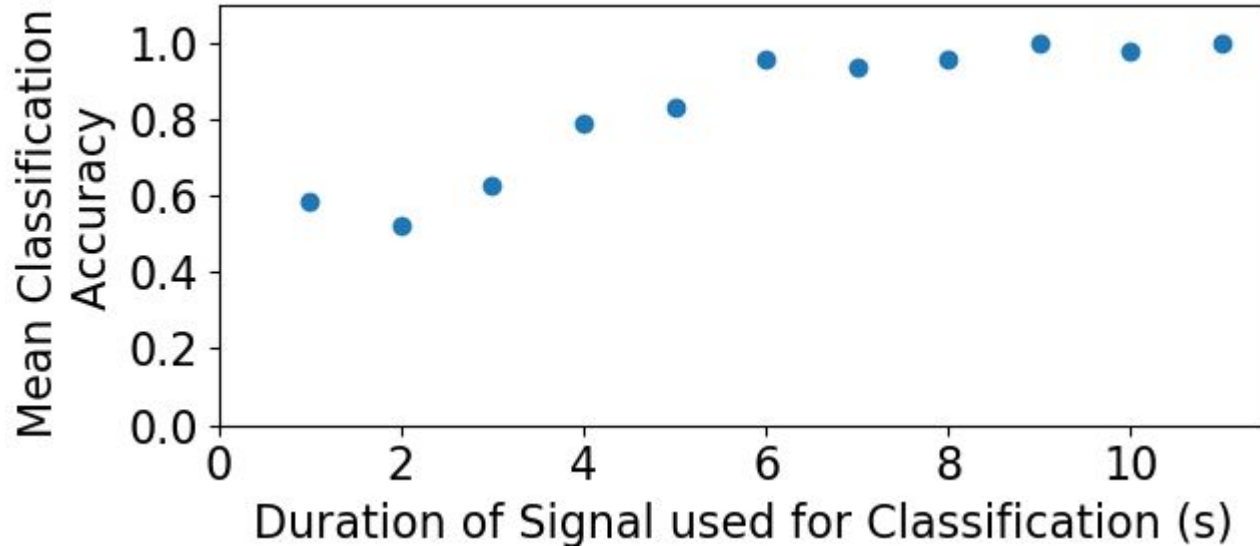
Duration of Human Input- black dotted line





# SSVEP: Performance Considerations

- Stimuli duration
- Number of options



# Summary

Hierarchical system for shared control of a humanoid robot.

- Shared Autonomy
  - Leverages the strengths of both humans and robots.
  - Reduces BCI Fatigue
- Hierarchical and configurable
- Intuitive screen-based visualization of the task
  - Enhances operator understanding and interaction.
  - Web-based System (RoboWebTools); platform-agnostic
- Robust Assistive Robotics Application
  - Reliable BCI with SSVEP
- Benchmark Experimental Setup for Evaluation of BCI Systems

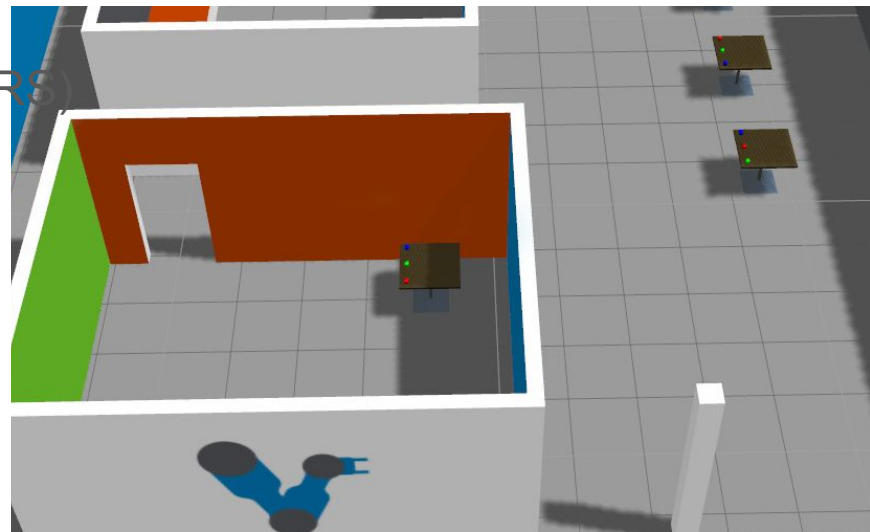
# Current/Next Steps

## Benchmark Evaluation of BCI-Robotic Systems

- Simulated Robotic environment
- Compare performance of different BCI Modalities
- SSVEP versus Eye-Tracking
- Hybridize BCI Modalities (SSVEP + fNIRS)

## BCI Robot Learning

- Interactive Robot Learning using BCI



Questions?