



# Remote Teleoperation Data Collection

B.S.: Aaron Zheng, Akshaj Gupta, Kourosh  
Salahi, Ziteng (Ender) Ji  
M.S.: Samuel Mankoff

ME206A Project Group 18

# Why data collection is important?

Open X-Embodiment: Robotic Learning Datasets and RT-X Models

Open X-Embodiment Collaboration  
(Hover to display full author list)



DROID: A Large-Scale In-the-Wild Robot Manipulation Dataset

DROID Dataset Team  
(Hover to display full author list)

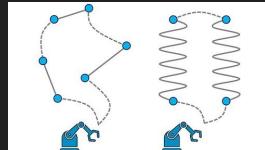
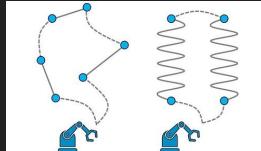
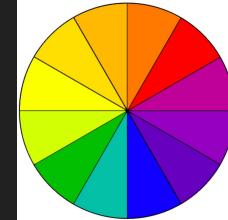
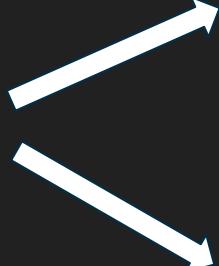
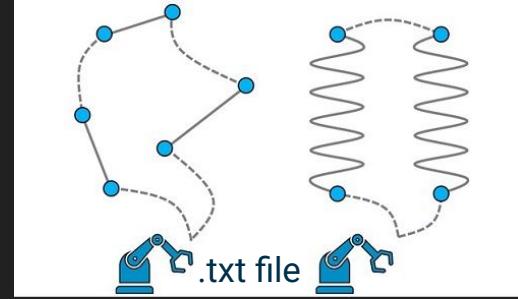


- Modern robot learning is **data-hungry**: powerful policies including imitation learning and diffusion models only perform as well as the demonstrations and interactions they're trained on.
- Large-scale datasets like **Open X-Embodiment** and **DROID** demonstrate the critical importance of real-world robot data by pooling demonstrations from many research labs and diverse robotic platforms.
- **Our contribution:** We introduce a novel approach to collect real-world manipulation data through intuitive teleoperation using Meta Quest 3 VR and hand controllers.

# Project Pipeline



+



# Project Components



## VR-Based Control (Planning)

Enable intuitive control of the UR7e robotic arm using Meta Quest 3 headset and controllers for natural 6-DoF manipulation.



## Data Collection & Storage

Capture high-fidelity trajectory data during teleoperation sessions and save demonstrations for training robot learning models.



## Trajectory Replay (Actuation)

Replay recorded trajectories by loading saved text files and executing the stored motion commands on the physical robot.



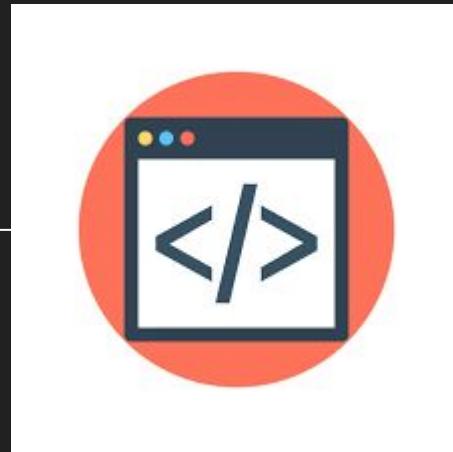
## Vision-Based Selection Sorting (Sensing)

Intelligently select and replay specific trajectories based on visual input from a color sensor for conditional behavior.

# VR Based Control



SteamVR,  
ALVR

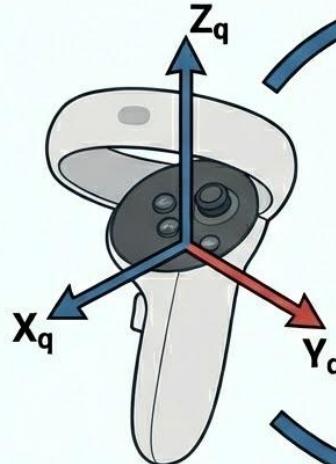


replay,  
ur7e\_coms  
moveit  
ros2



# VR Based Control

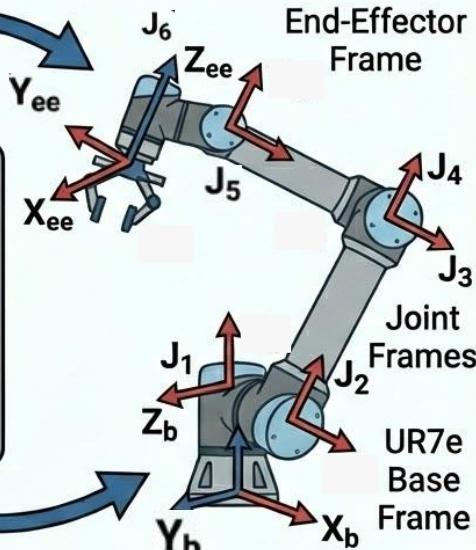
## Coordinate Frame Mapping & Pose Transformation



$$T_{\text{target}} = \underbrace{T_{\text{robot}}^{\text{home}} \cdot (T_{\text{vr}}^{\text{calibration}})^{-1}}_{\text{Calibration Offset}} \cdot T_{\text{vr}}^{\text{live}}$$

Pose  
Transformation &  
Inverse Kinematics

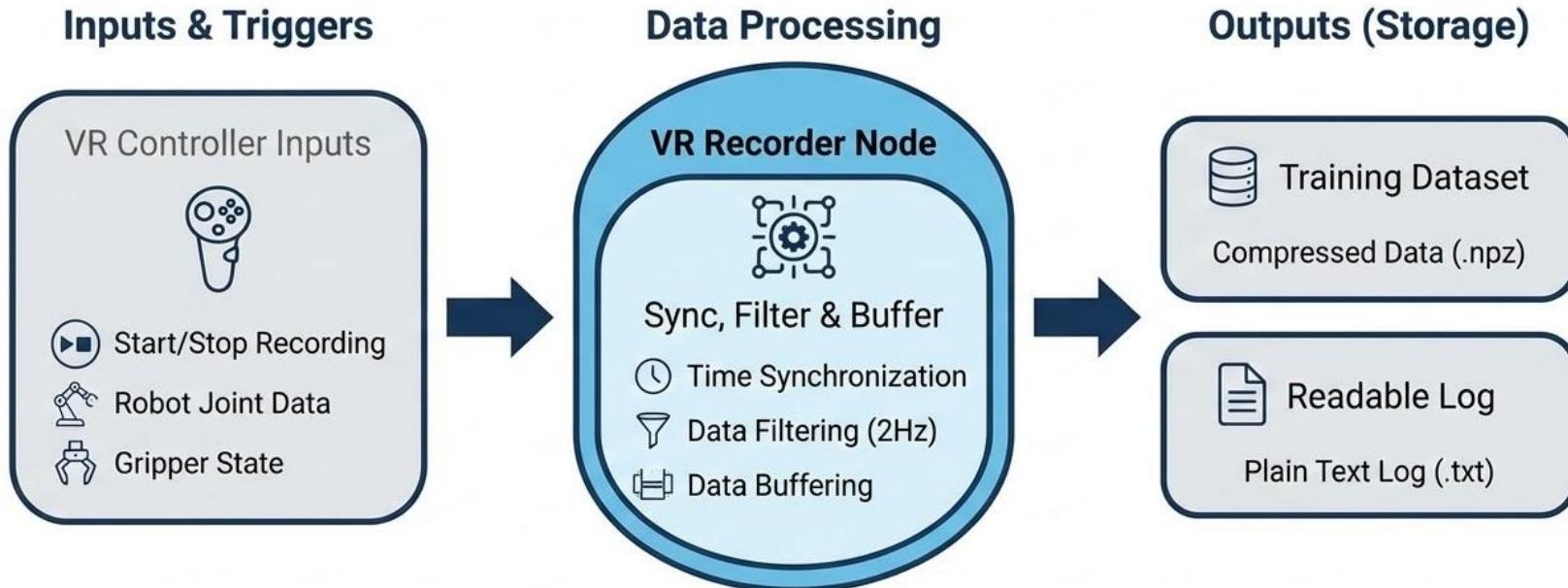
Quest 3  
Controller Frame



Joint Angles:  
[ $\theta_1, \theta_2, \theta_3, \theta_4, \theta_5, \theta_6$ ]

# VR Data Recording Pipeline

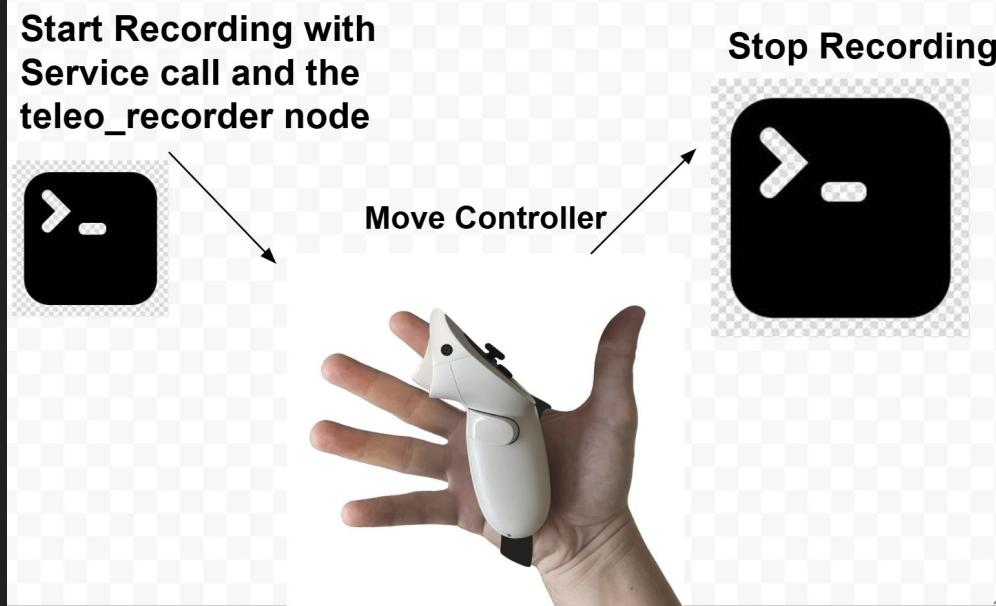
**Data Flow:** Inputs → Processing → Storage



# VR Recording & Saving

VR recorder, detailed steps:

- 1) Run teleo\_recorder node
- 2) Run the corresponding service call (/start\_recording) to start
- 3) While performing teleoperation, this service will sample ur7e joint state at 2Hz (with timer)
  - a) Samples every 0.5s for smooth replay
- 4) Run (/stop\_recording) to end and save joint states + gripper states in .txt file



# Sensing & Replay

'blue', 'green', 'none'

detected\_color



Ur7e\_Executor

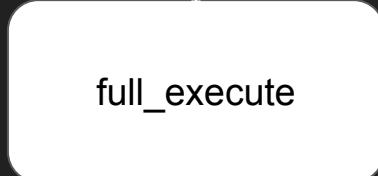
camera\_detection

full\_execute

'blue', 'green'

Populate  
Job Queue

Replay  
Trajectory



# Biggest Challenge

- Hardware
  - Connecting Quest headset and controller to the same local network as the lab computer
- Software
  - Smooth trajectory replay; aligning recording sampling rate and replay speed



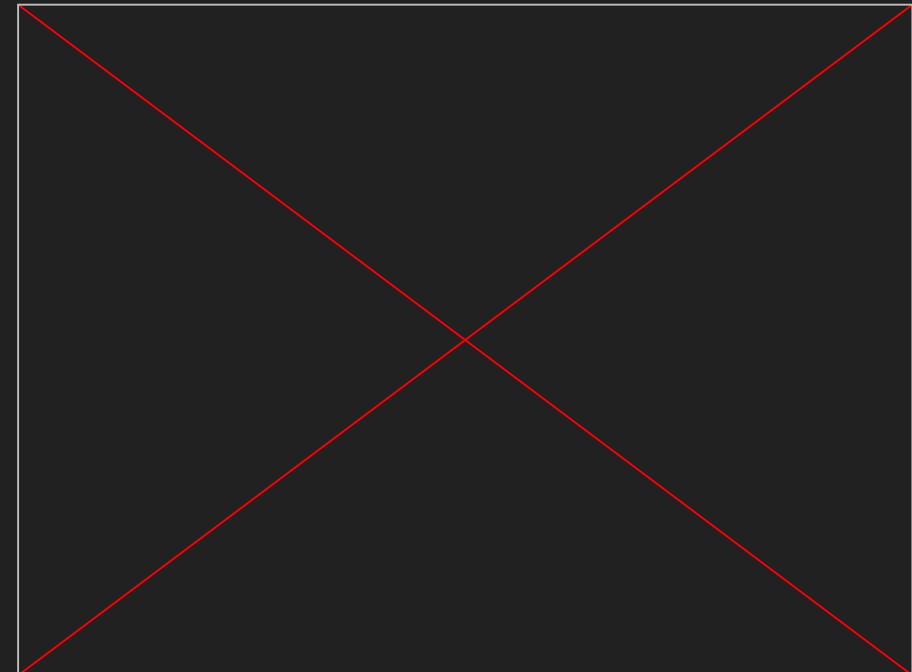
# Future Considerations

## Controls

- We experimented with controls
  - PID controls per joint
  - Speed limiting
  - Reduces noise
- Additionally safety Robot bounding box
  - UR7e Proximity



# Video Demo



# Video Demo

