Welcome

Movelt Pro Works Here Webinar Series





Madison
Clark-Turner, PhD
Robotics Scientist
Charles River Analytics



Nikolaas
Bender
Robotics Software Engineer
Charles River Analytics



Michael
Giancola, PhD
Al Scientist
Charles River Analytics



Nathan
Brooks, PhD
CTO
PickNik Robotics



Dave Grant
CEO
PickNik Robotics



Housekeeping for Today



More about PickNik Robotics



We want to hear from you! Please drop your questions in the Q&A tab!



More about Charles River Analytics





Mission: Empower organizations to deploy robots of all shapes and sizes with human-like intelligence to elevate productivity and enrich lives



Robotics Application Platform



Concept & Development

- Al Training
- Rapid Development
- Digital Prototype

Pilot & Scale Up

- Deploy & Run Apps
- Adapt to Evolving Use Cases
- Add Enhancements







Madison Clark-Turner, PhD
Robotics Scientist

- Leads a portfolio focused on human roles in human-in-the-loop operations
- Work spans capturing nonverbal cues, reconstructing casualties for triage, and enhancing situational awareness
- PI for NASA's <u>Person Aware Liaison</u> (<u>PAL</u>), building collaborative robotics that integrate human inputs and foster trust in AI



Nikolaas Bender Robotics Software Engineer

- Specializes in reinforcement learning for robotic arm control
- Develops and integrates perception systems and deep learning for image analysis
- Plays a key role in advancing the capabilities of the <u>Awarion®</u> <u>Autonomous Lookout System</u> through applied R&D and extensive testing



Michael Giancola, PhD Al Scientist

- Expert in natural language processing (NLP), large language models, and automated reasoning methods
- Applies deep learning for speech understanding and information extraction on the PAL system
- Developed a novel algorithm for theory-of-mind reasoning under uncertainty



Intro

- Charles River Analytics develops advanced robotics and autonomy solutions that combine adaptive AI, leading-edge perception, and intuitive human-machine interaction.
- We specialize in building modular autonomy components that integrate seamlessly with current and emerging robotic platforms.
- Our teams draw on deep expertise in machine learning, computer vision, cognitive science, and human factors to create systems that operate effectively across ground, air, maritime, and space environments.
- From reinforcement learning for manipulation to multimodal sensing for situational awareness, our work enables robots to act safely, intelligently, and collaboratively in dynamic, real-world settings. Whether supporting complex missions or enabling resilient autonomy in the field, we're focused on delivering robust, explainable, and readyto-deploy robotic solutions.



















- Robots are being placed alongside humans in many settings
 - Especially in unstructured environments
- Robots and humans have roles they excel in and make good teams...
 - ...in theory







How are robots currently perceived?



Dumb

- Tools, not partners
- Laborious control
- Structured tasking



Frustrating

- Fail to meet expectations
- Require restatement of commands

Untrustworthy



Unsafe

- Unaware of humans
- Unable to respond to anomalies



So How Do We Build Trust?







- In humans trust is built through a combination of factors:
 - Expectation versus capability
 - Communication
- Humans have an expectation of peers when collaborating
- If expectations are higher than capability, we communicate our limitations
 - Spoken words
 - Gesture
 - Engagement
- What we need is...
 - A human-to-robot translator







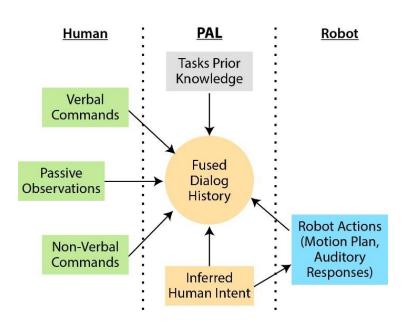
- Multimodal user input facilitates holistic user understanding
- Enhanced contextual understanding allows the robot to respond intelligently to prompts
- Communicates bidirectionally to ask clarifying questions
- Capable











The generalized PAL pipeline

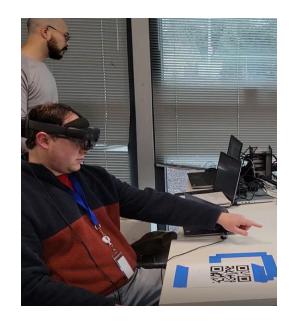
The human said "can you pick that up and bring it here" The human user is looking at "screw" The human user is pointing at "toolbox" PAL registers that the human wants the "screwdriver" The robot attempts to grasp the screwdriver PAL identifies an obstruction The robot said "I have picked up the screwdriver but I am unable to bring it to you due to an obstruction" The human user is looking at "toolbox" The human user said "Ok, I will open that in a minute" PAL determines that the human is busy and will wait

An example interaction transcript











Speech

Point/Gaze



Observations





no visible landm

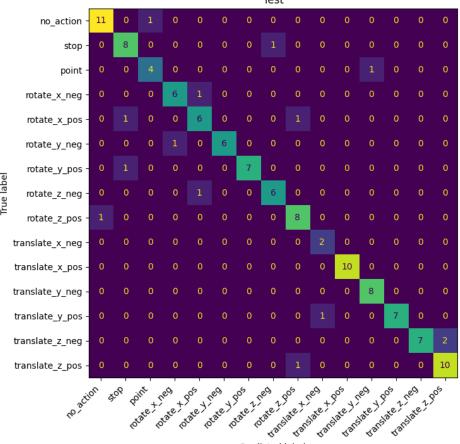
Gesture

Body Posture



Observations (Gesture Accuracy)





Predicted label

14

Λ

- 10

Inference



User Observations **SPEECH:** "Can you pick up that pipe and

bring it to me"

POINT: Pointing at Pipe #3

GAZE: Looking at Pipe #3

LLM

Robot Instructions ACTION: pick_up_and_bring ACTION_TARGET: pipe #3 DESTINATION: user's hand

Large Language Model

- Consumes interaction transcript and generates robot-interpretable output
- Avoids grammar dependency
- Can be easily modified to incorporate new modalities and higher-level insights



Robotic Control (PickNik)

ACTION: pick_up_and_bring ACTION_TARGET: pipe #3 DESTINATION: user's hand





Factors we considered when choosing Movelt Pro

- Platform agnosticism
- Advanced grasp behaviors
- Easy to develop and implement robotics behaviors
- Movelt is well known for work with robot manipulators

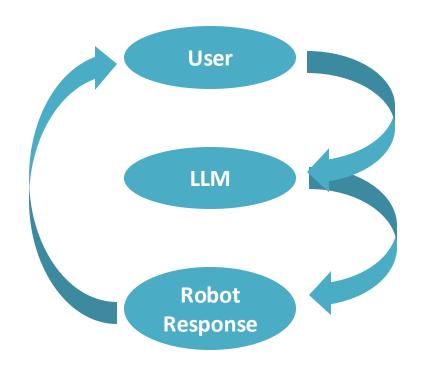


charles river

analytics

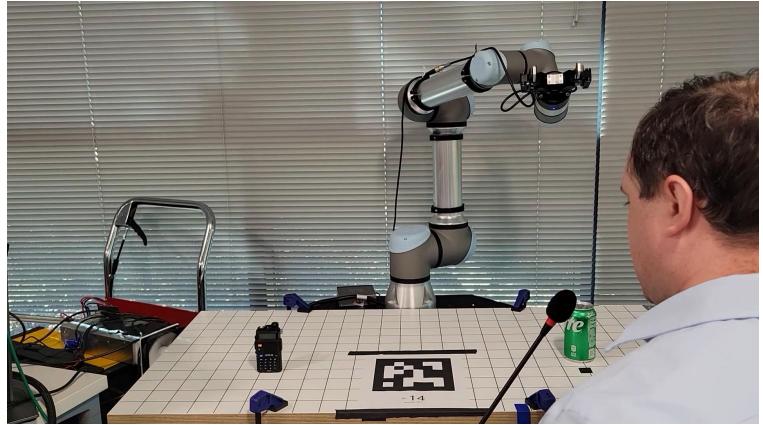
Ongoing Interaction

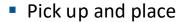






What We Can Do



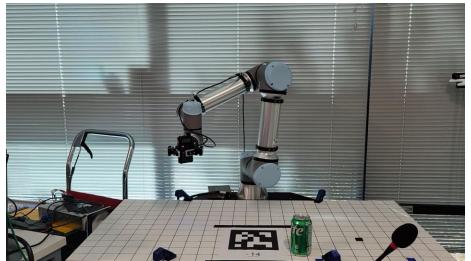


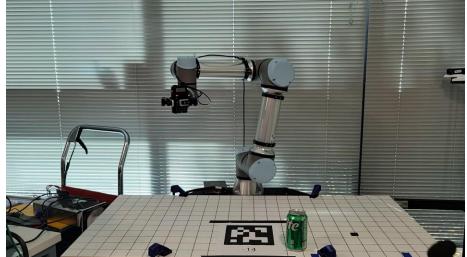
Manipulate (translate and rotate)

















- Temporal context
 - "Put the screwdriver back where you found it"
- Spatial context
 - "Pick up the can on the right"
- Descriptive context
 - "Pick up the blue cup"
- Task context
 - While looking at a screwdriver: "Get me a fastener for this tool"



PAL Applications



Assistive care robotics



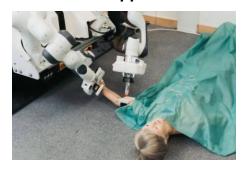
Manufacturing and repair



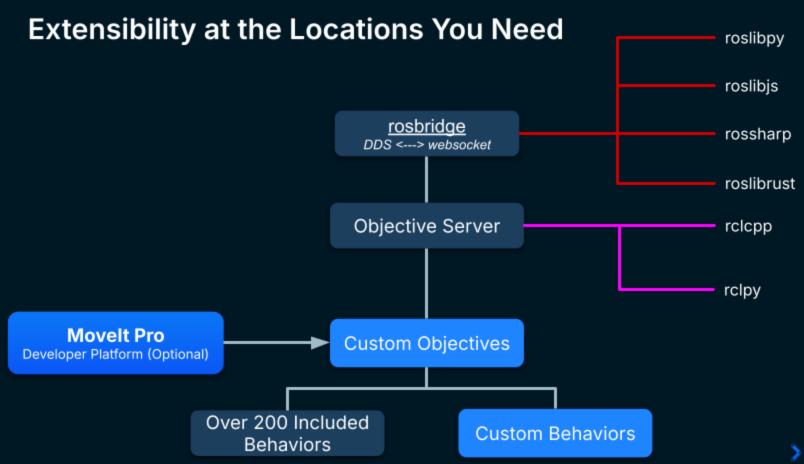
Vehicle-level autonomy

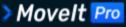


Medical applications









CPU, GPU, and Arm Agnostic

- We package everything in Docker to make machine provisioning easy and new developer spin-up painless.
- Our software runs on x86 and ARM architecture.
- We provide GPU agnostic ONNX models.
 - Our Docker containers pass through CUDA acceleration, including on <u>Nvidia Jetson devices</u>.
- We maintain and expand a growing list of ROS compatible hardware and driver ratings.



A Framework for Using ML Reliably & Safely









Cycle Time Planners

Input

Motion request

Output

Joint trajectory

Example

Cycle time optimal motion between work areas.

Segmentation Transformers

Vision (2D or 3D), Text Prompt

Mask specifying location

Find the next component for your task.

Grasp Transformers

Vision (2D or 3D)

Grasp pose

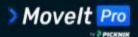
Grasp novel or deformable objects.

Action Policies

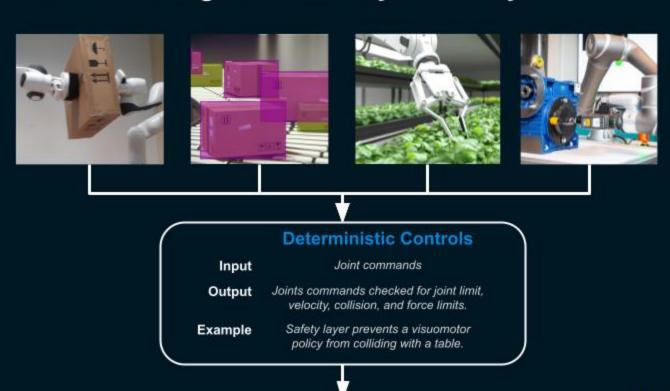
Vision (2D or 3D)

Joint trajectory

Plan contact-rich motion for assembly or manipulation.

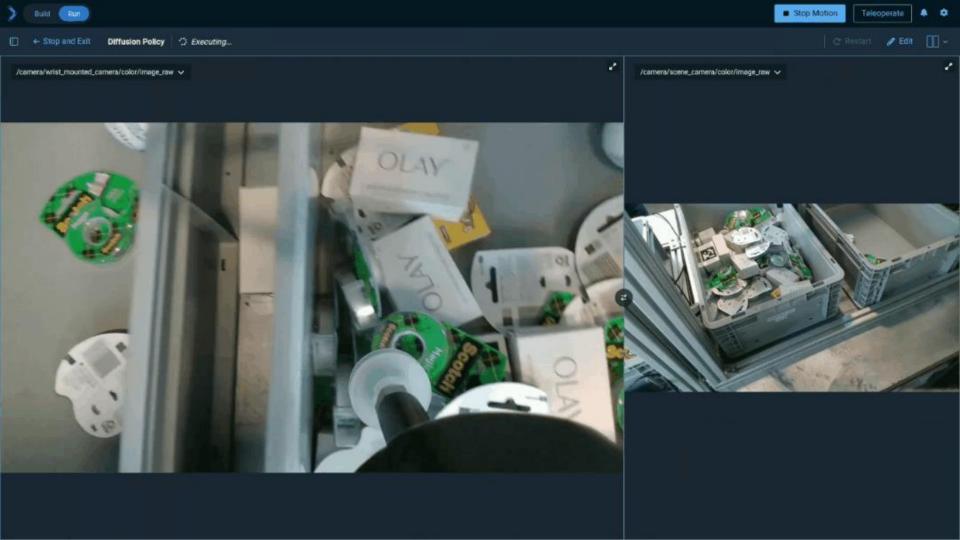


A Framework for Using ML Reliably & Safely





Robot Hardware or Simulation



Questions for our Presenters?

Movelt Pro Works Here Webinar Series





Madison
Clark-Turner, PhD
Robotics Scientist
Charles River Analytics



Nikolaas
Bender
Robotics Software Engineer
Charles River Analytics



Michael
Giancola, PhD
Al Scientist
Charles River Analytics



Nathan
Brooks, PhD
CTO
PickNik Robotics



Dave Grant
CEO
PickNik Robotics



Questions for our Presenters

More Success Stories: https://picknik.ai/case-studies/
BMW & Hivebotics Webinar recordings: https://picknik.ai/webinars/









CleanBotix

Don't Forget to Join us on
Thursday, October 30th at 12p EST
Featuring Adam Jacques of Clean-Botix,
A Food Sanitation Company

Registration Link is in the chat!





Questions for our Presenters?

Movelt Pro Works Here Webinar Series





Madison
Clark-Turner, PhD
Robotics Scientist
Charles River Analytics



Nikolaas
Bender
Robotics Software Engineer
Charles River Analytics



Michael
Giancola, PhD
Al Scientist
Charles River Analytics



Nathan
Brooks, PhD
CTO
PickNik Robotics



Dave Grant
CEO
PickNik Robotics

