



Background and Motivation

- Ground robots commonly used in **autonomous exploration** of extraterrestrial or unknown environments
 - Typical rover designs have trouble climbing vertical steps
 - Robots designed specifically to climb are complex, expensive, and impractical for most applications



(a) NASA's Curiosity

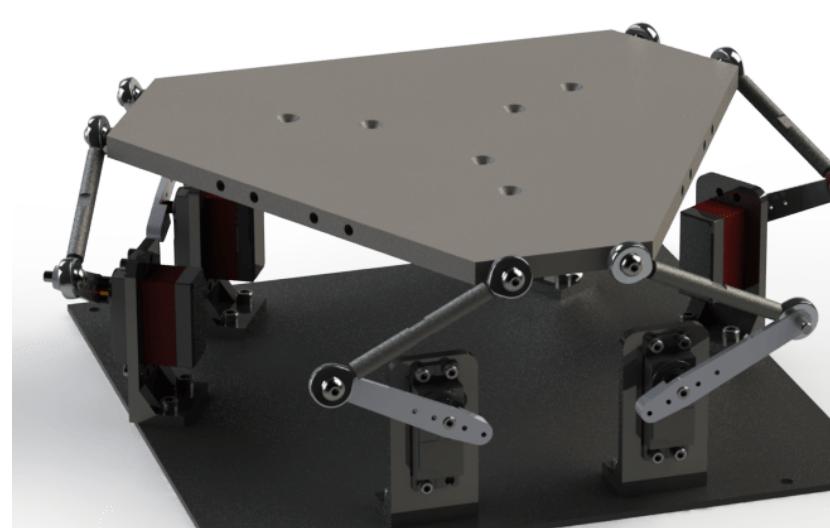


(b) Stair Climbing Robot

- Sensitive instruments** often need to be transported while traversing difficult terrain
 - Few robots have active stabilization systems implemented to counteract the effect of the environment
- We propose a low-cost system that implements a passive suspension system for driving over rough terrain and an active stabilized platform for protecting its payload
 - Optimized **rocker-bogie wheelbase** with six driven wheels for climbing over large obstacles
 - Actively stabilized platform** for reducing forces on the payload
 - Autonomous control of robot using position and orientation data from indoor motion capture system
 - Autonomous navigation using **path planning algorithms** and depth measurements of the surrounding environment

Platform

- A 6DoF **Fixed Rotary Stewart Platform** was designed
- The kinematics and dynamics of the system were tested using MATLAB and the Simscape Multibody plugin
- High torque metal gear hobby servos were used to provide adequate force and a fast response time for quickly overcoming disturbances
- All mounts, brackets, and arms were **3D printed** using PLA plastic; steel ball joints were used for smooth actuation of joints



(a) Platform Render



(b) Constructed Platform

- A 9-axis inertial measurement unit (IMU) mounted on the actuated platform provides orientation and acceleration data for control

Platform Control

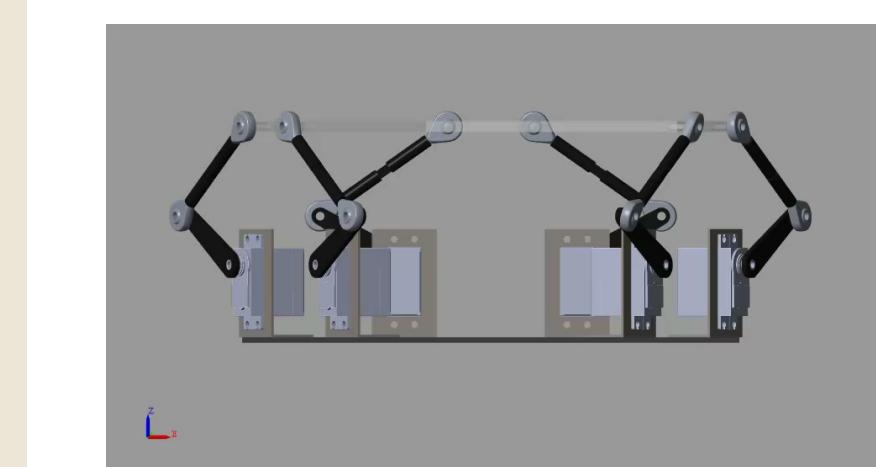
- Nonlinear control law** accounts for platform orientation, translation, and acceleration:

$$x_n(t) = A_n \sin(\alpha(t)) + B_n \sin(\beta(t)) + C_n \sin(\gamma(t)) + D_n x(t) + E_n y(t) + F_n z(t)$$

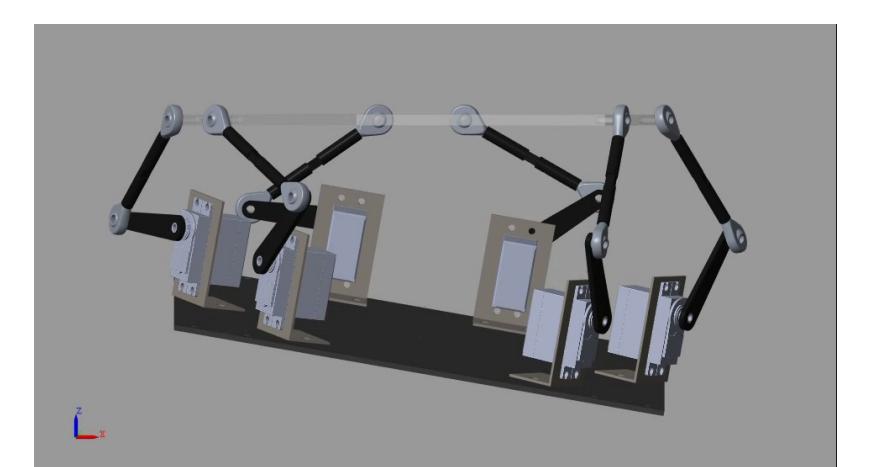
$$d\theta_n(t) = x_n(t) + G_0 \int_0^\infty \zeta(t) x_n(t) dt + H \ddot{x}_n(t)$$

$$\theta_n(t) = \theta_n(t - t_o) + d\theta_n(t) \Delta t$$

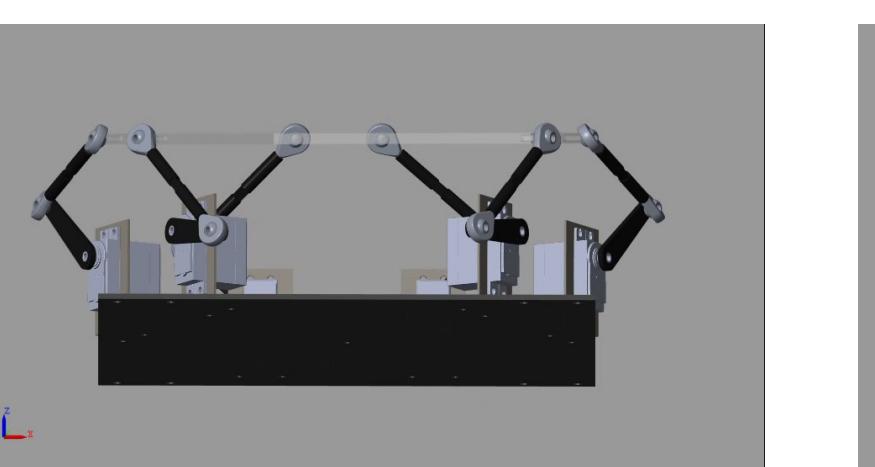
- Simscape Multibody** simulation with implemented controller
 - Platform base was subjected to rolling and pitching in a Lissajous motion with maximum inclinations of 20°
 - The controller was instructed to maintain a platform roll and pitch of 0° while minimizing yaw and translational motions



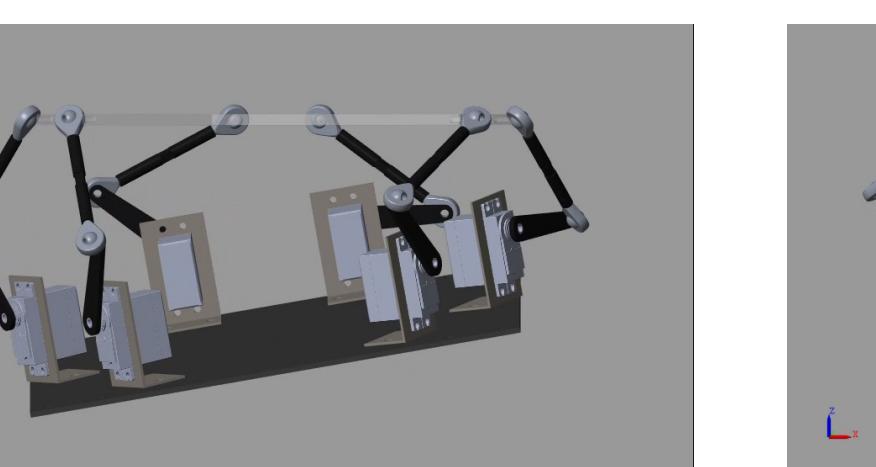
(a) 0 sec



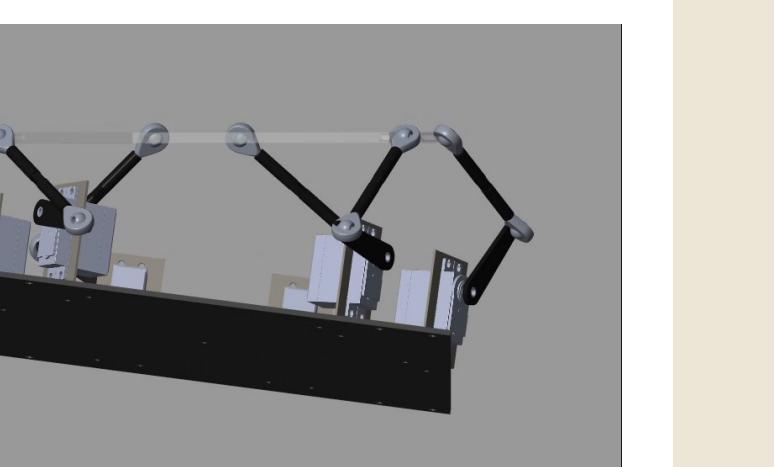
(b) 5 sec



(c) 10 sec



(d) 15 sec



(e) 20 sec

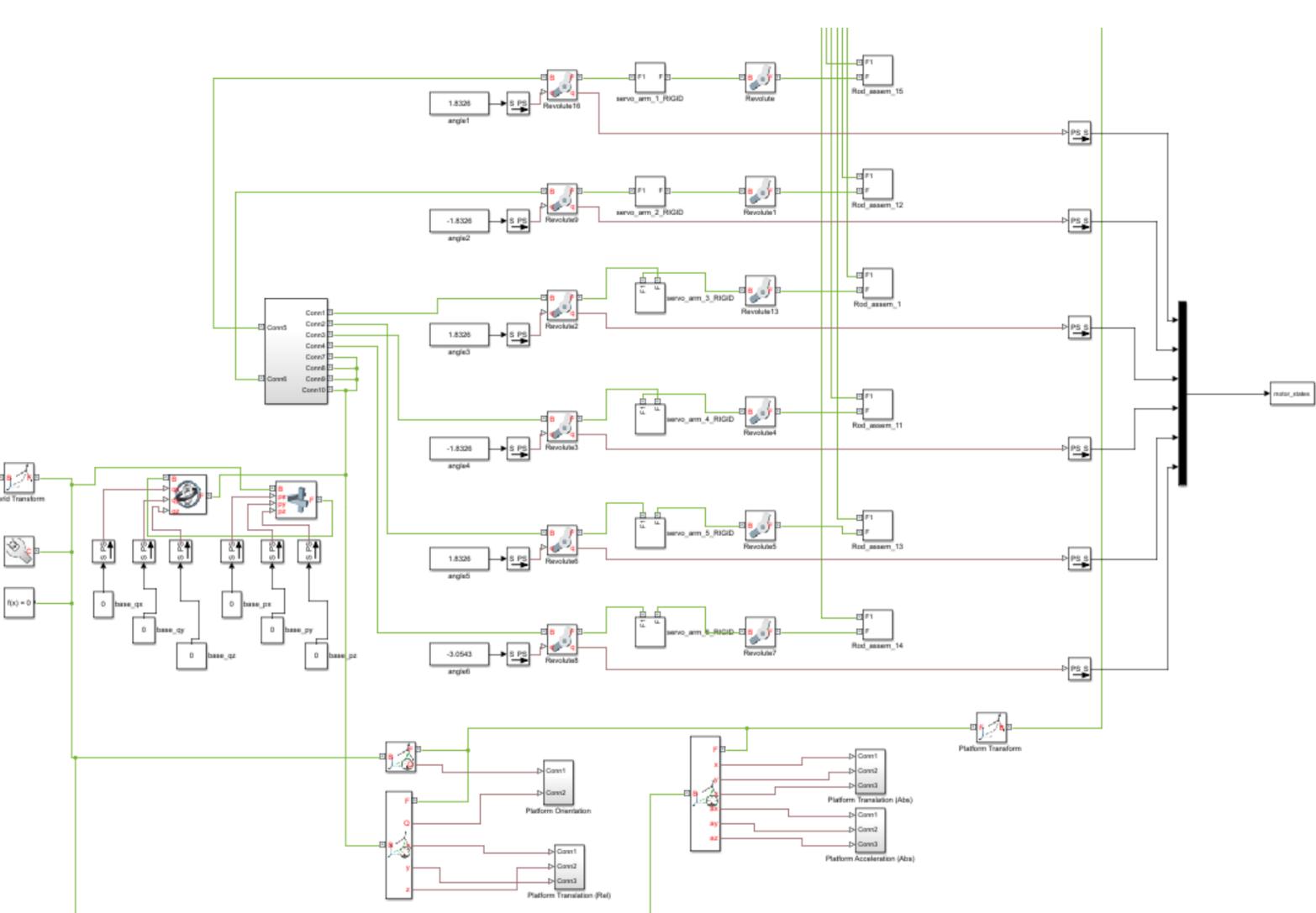
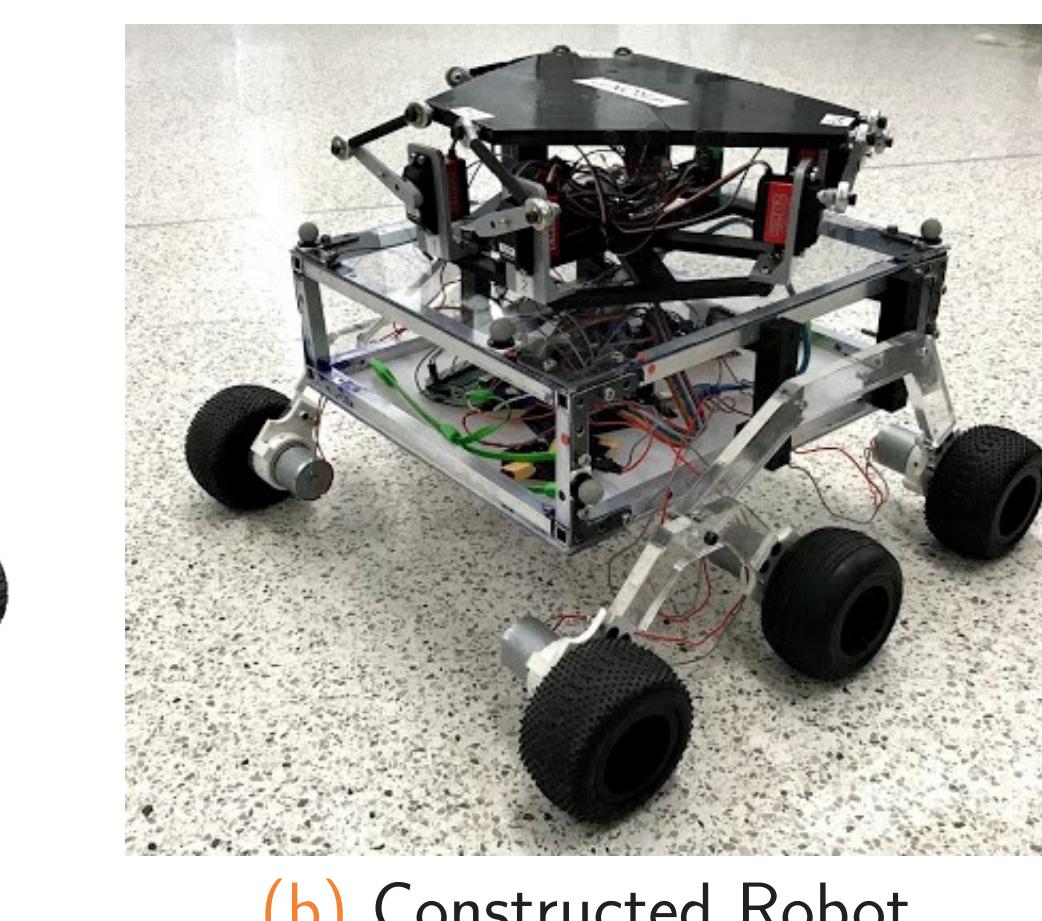


Figure: Simulink Model for Platform

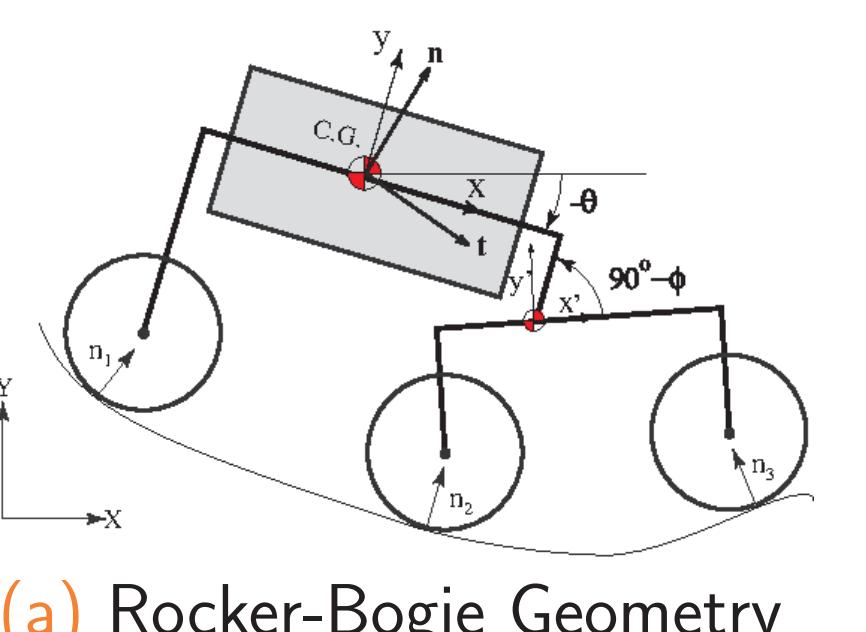
Robot

- NVIDIA Jetson TX2** embedded GPU runs all control and path planning algorithms
- Servo and motor commands sent over hardware serial ports to Arduino Megas

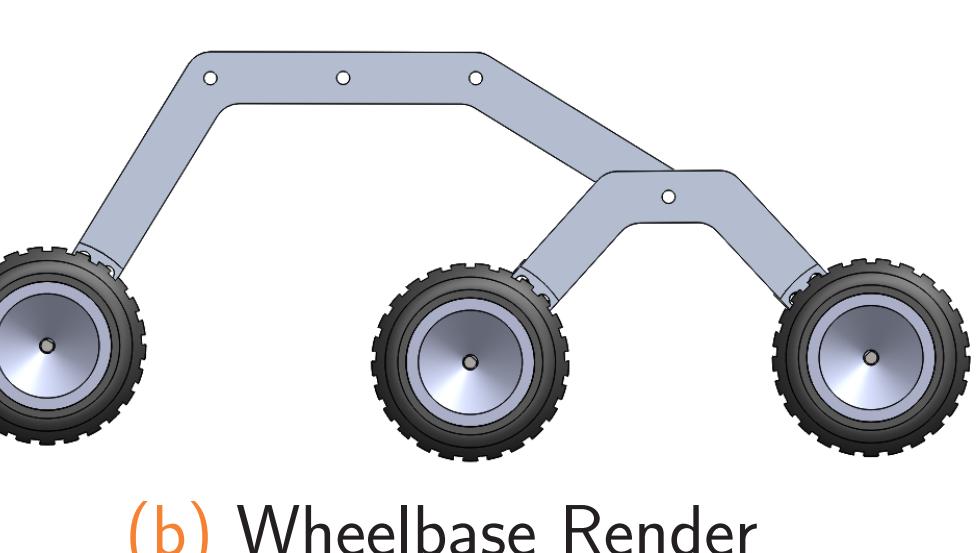


(a) Robot Render

- Modular design** allows for simple system alterations
- Lightweight yet durable aluminum structure
- 3D printed components reduce manufacturing time and allow for rapid testing of prototypes
- Six-wheeled **Rocker-Bogie** system allows for climbing a greater step size



(a) Rocker-Bogie Geometry



(b) Wheelbase Render

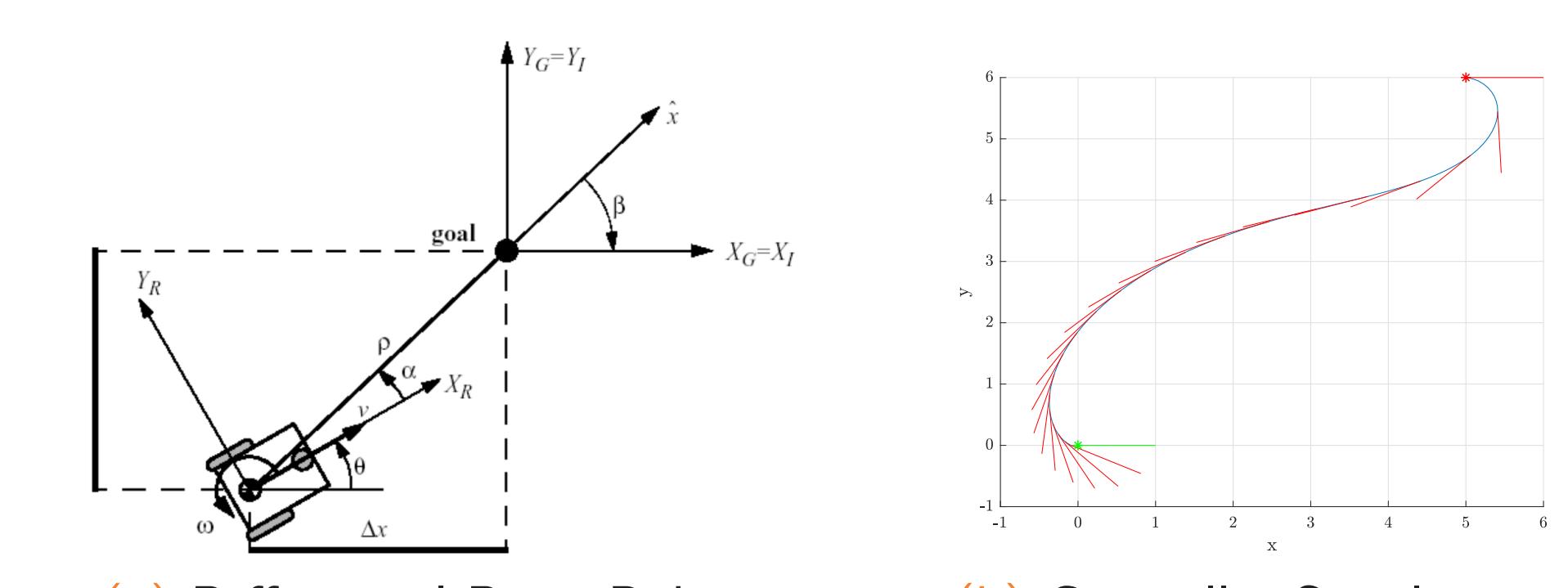
Robot Control

- Differential drive robot** model assumed for simplicity
- Using geometric relationships between the robot's position and orientation and the goal, a control law can be developed:

$$\rho = \sqrt{\Delta x^2 + \Delta y^2}, \alpha = -\theta + \text{atan}2(\Delta y, \Delta x), \beta = -\theta - \alpha + \theta_g$$

$$v = k_\rho \rho, \omega = k_\alpha \alpha + k_\beta \beta$$

- Able to control robot to drive to desired position and orientation in global coordinates or to a series of way-points to follow a more **complex trajectory**



- Tuning the controller gains (k_ρ , k_α , and k_β) varies the shape of the robot's trajectory

Autonomous Navigation

- Path planning system uses depth and localization data to navigate around obstacles towards its goal

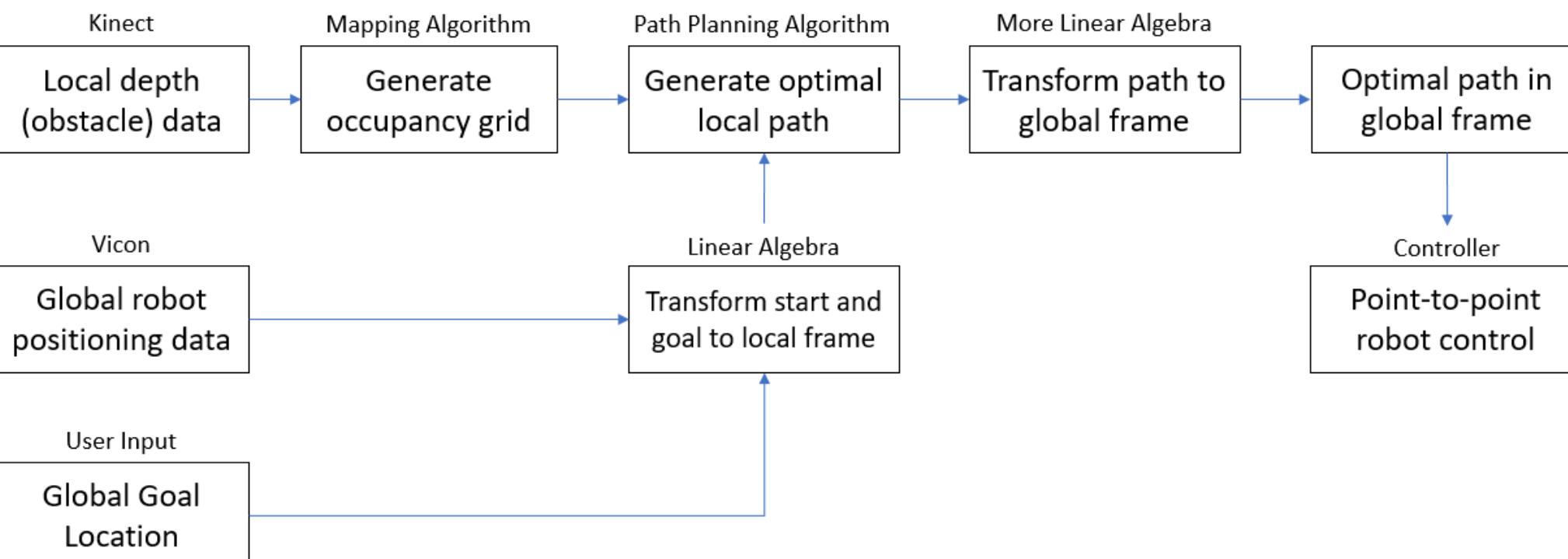
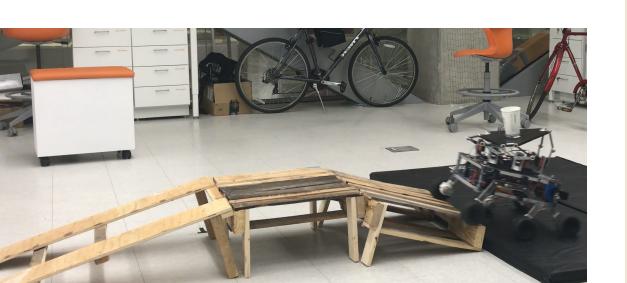
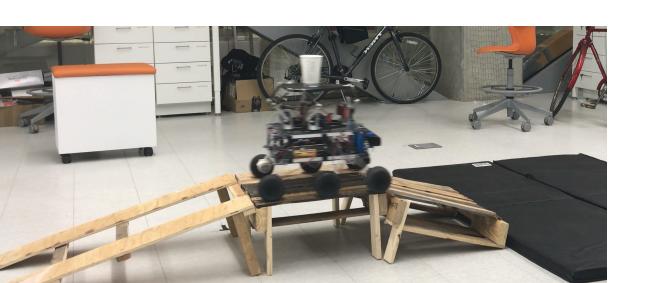


Figure: Navigation System Block Diagram

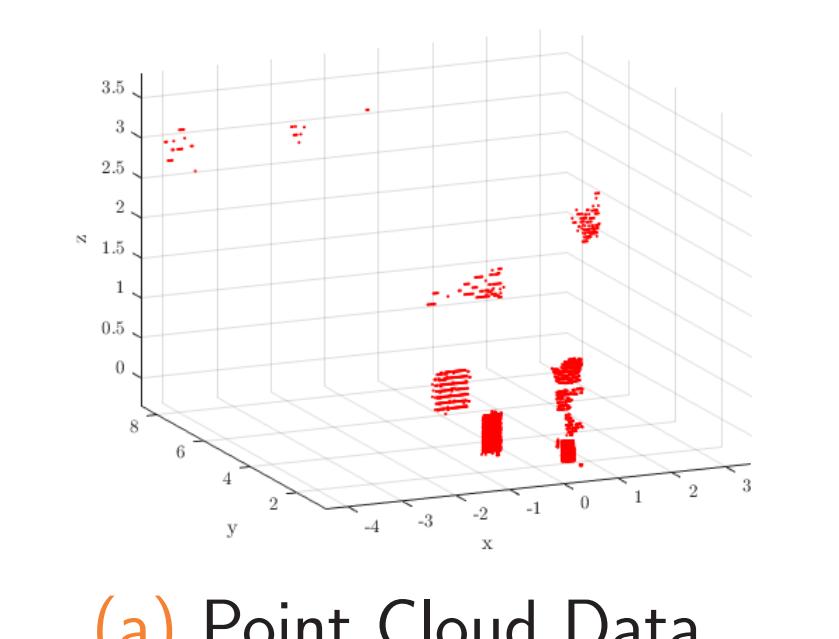
- Vicon motion capture system** provides data on robot's global position
- Microsoft Kinect** provides local data on obstacle locations using a **3D point cloud** of the surrounding environment
- Point cloud data converted to **2D occupancy grid** based on obstacle characteristics (height, gradient, etc.)
- Dijkstra's algorithm** calculates minimum cost path to goal in the local frame and is transformed into global frame for point-to-point navigation

Experimental Results and Conclusion

- Incline test



- Obstacle avoidance test



- Was able to successfully traverse through and over obstacles while protecting a payload on the platform
- Designed autonomous navigation, control, and obstacle avoidance system that significantly exceeds project requirement scope
- Project exceeded all customer design requirements and was overall a great success