

Implementation and Control of an Underactuated Ballbot System

Technical Presentation

Robotics Engineering

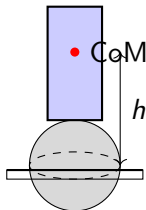
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Outline

- 1 Introduction to Ballbot Concept
- 2 Mechanical Architecture
- 3 Hardware Architecture
- 4 System Modeling
- 5 Dynamic Equations
- 6 Omni Kinematics
- 7 Control System Design
- 8 Software Architecture
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- 10 Stability and Tuning
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What is a Ballbot?

- Dynamically stable robot
- Balances on a single spherical ball
- Inherently unstable (inverted pendulum)
- Omnidirectional mobility
- Underactuated MIMO system



Key Challenges

Underactuated System

3 actuators (motors) controlling 5 degrees of freedom:

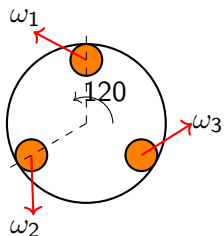
- Body tilt (2 DOF): θ_x, θ_y
- Body rotation (1 DOF): ψ
- Ball position (2 DOF): x, y

Control Requirements

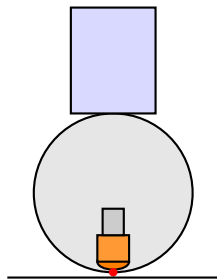
- Fast sensor sampling (> 100 Hz)
- Low-latency control loop
- Robust to disturbances
- Smooth omnidirectional motion

Three-Roller Internal Drive

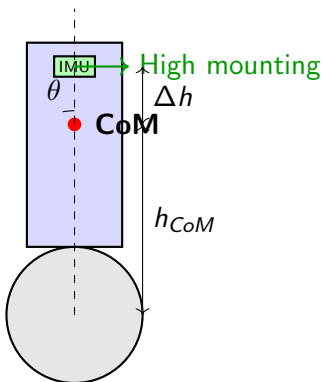
Top View (120° Configuration)



Side View



Center of Mass and IMU Placement

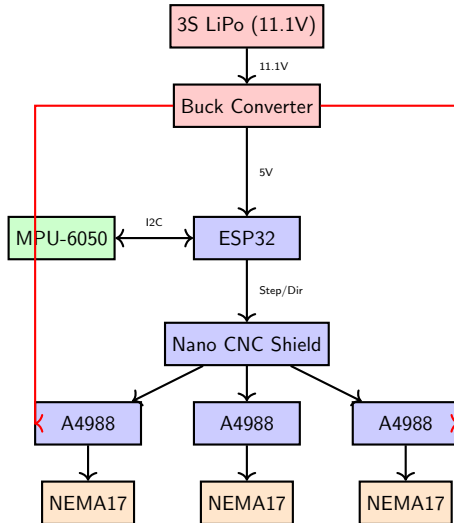


Design Criteria:

- CoM high above ball center
- Increased $h \rightarrow$ longer time constant
- IMU near top for better sensitivity
- Symmetric mass distribution

$$T_{fall} \propto \sqrt{\frac{h}{g}}$$

Electronics Block Diagram



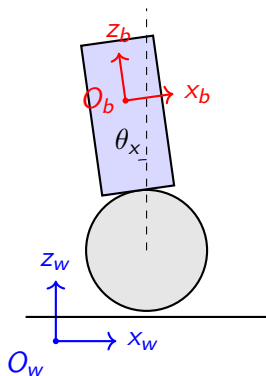
Component Specifications

Component	Specification	Purpose
ESP32	240 MHz dual-core	Control + sensor fusion
MPU-6050	6-axis IMU, 1 kHz max	Tilt measurement
A4988	2A max, 1/16 microstep	Stepper driver
NEMA17	1.5A, 0.4 Nm	Roller actuation
3S LiPo	11.1V, 2200 mAh	Power supply
Buck Conv.	11.1V \rightarrow 5V, 3A	Voltage regulation

Key Design Choices

- ESP32: Sufficient speed for real-time control (< 10 ms loop)
- MPU-6050: Cost-effective 6-DOF with reasonable noise
- A4988: Simple step/dir interface, adequate current

Coordinate Frames



Frames:

- World frame: $\{O_w\}$
- Body frame: $\{O_b\}$
- Ball frame: $\{O_{ball}\}$

Tilt Angles:

θ_x : roll (about x_w)

θ_y : pitch (about y_w)

Small Angle Assumption:

$$|\theta_x|, |\theta_y| < 15$$

State and Input Vectors

State Vector

$$\mathbf{x} = \begin{bmatrix} \theta_x \\ \theta_y \\ \dot{\theta}_x \\ \dot{\theta}_y \end{bmatrix} \in \mathbb{R}^4$$

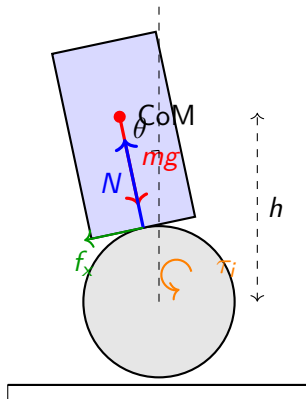
- θ_x, θ_y : Tilt angles (rad)
- $\dot{\theta}_x, \dot{\theta}_y$: Angular velocities (rad/s)

Input Vector

$$\mathbf{u} = \begin{bmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{bmatrix} \in \mathbb{R}^3$$

- τ_i : Torque from motor i (Nm)

Free Body Diagram



Inverted Pendulum Analogy

Linearized Dynamics:

$$I\ddot{\theta} = mgh \sin \theta - \tau_{motor}$$

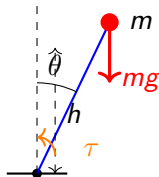
$$\approx mgh\theta - \tau_{motor}$$

For small angles:

$$\ddot{\theta} = \frac{mgh}{I}\theta - \frac{\tau_{motor}}{I}$$

Define: $\omega_n^2 = \frac{mgh}{I}$ (natural frequency)

$$\ddot{\theta} = \omega_n^2\theta - \frac{\tau_{motor}}{I}$$



Torque to Angular Acceleration

Motor torques produce body angular acceleration:

$$\begin{bmatrix} \ddot{\theta}_x \\ \ddot{\theta}_y \end{bmatrix} = \begin{bmatrix} \omega_n^2 & 0 \\ 0 & \omega_n^2 \end{bmatrix} \begin{bmatrix} \theta_x \\ \theta_y \end{bmatrix} + \frac{1}{I} \mathbf{J}_{motor} \begin{bmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{bmatrix} \quad (1)$$

where \mathbf{J}_{motor} maps motor torques to body angular accelerations.

For 120° configuration:

$$\mathbf{J}_{motor} = \frac{r_{ball}}{3} \begin{bmatrix} 0 & -\sqrt{3}/2 & \sqrt{3}/2 \\ 1 & -1/2 & -1/2 \end{bmatrix}$$

where r_{ball} is the ball radius.

Tilt to Translation Relationship

Controlled tilt produces ball motion:

$$\ddot{\mathbf{p}}_{ball} = g \begin{bmatrix} \tan \theta_y \\ -\tan \theta_x \end{bmatrix} \approx g \begin{bmatrix} \theta_y \\ -\theta_x \end{bmatrix}$$

Principle:

- Tilt forward ($\theta_y > 0$) \rightarrow ball rolls forward
- Tilt right ($\theta_x > 0$) \rightarrow ball rolls right
- Acceleration proportional to tilt angle

Control Strategy

To move in direction $\mathbf{v}_{desired}$:

- 1 Command tilt: $\theta_{cmd} = k_v \mathbf{v}_{desired}$
- 2 Maintain tilt via motor torques
- 3 Ball accelerates in desired direction

Simplified Nonlinear Model

Complete MIMO underactuated dynamics:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u})$$

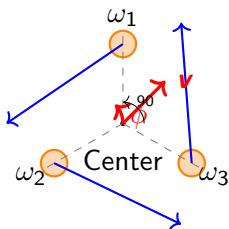
$$\begin{bmatrix} \dot{\theta}_x \\ \dot{\theta}_y \\ \ddot{\theta}_x \\ \ddot{\theta}_y \end{bmatrix} = \begin{bmatrix} \dot{\theta}_x \\ \dot{\theta}_y \\ \omega_n^2 \sin \theta_x - \frac{c_1 \tau_1 + c_2 \tau_2 + c_3 \tau_3}{I} \\ \omega_n^2 \sin \theta_y - \frac{d_1 \tau_1 + d_2 \tau_2 + d_3 \tau_3}{I} \end{bmatrix}$$

where c_i, d_i are kinematic coefficients from \mathbf{J}_{motor} .

Linearized (small angles):

$$\dot{\mathbf{x}} = \mathbf{Ax} + \mathbf{Bu}$$

Three-Wheel Kiwi Drive Kinematics



Velocity decomposition:

Global velocity:

$$\mathbf{v} = \begin{bmatrix} v_x \\ v_y \end{bmatrix}$$

Angular velocity: ω_r (rotation)

Goal: Find $[\omega_1, \omega_2, \omega_3]^T$ to achieve \mathbf{v} and ω_r

Inverse Kinematics Derivation

Velocity at wheel i perpendicular to mounting:

$$v_{\perp,i} = \mathbf{v} \cdot \mathbf{n}_i + r_{config}\omega_r$$

where \mathbf{n}_i is the unit normal vector for wheel i .

For 120° configuration:

$$\mathbf{n}_1 = \begin{bmatrix} -1 \\ 0 \end{bmatrix}, \quad \mathbf{n}_2 = \begin{bmatrix} 1/2 \\ -\sqrt{3}/2 \end{bmatrix}, \quad \mathbf{n}_3 = \begin{bmatrix} 1/2 \\ \sqrt{3}/2 \end{bmatrix}$$

Wheel angular velocity:

$$\omega_i = \frac{v_{\perp,i}}{r_{roller}}$$

Matrix Form

Inverse kinematics (body to wheels):

$$\begin{bmatrix} \omega_1 \\ \omega_2 \\ \omega_3 \end{bmatrix} = \frac{1}{r_{roller}} \begin{bmatrix} -1 & 0 & r_{config} \\ 1/2 & -\sqrt{3}/2 & r_{config} \\ 1/2 & \sqrt{3}/2 & r_{config} \end{bmatrix} \begin{bmatrix} v_x \\ v_y \\ \omega_r \end{bmatrix}$$

Compact form:

$$\omega_{wheels} = J^{-1} \mathbf{v}_{body}$$

Note: For ballbot, ω_r is typically set to 0 (no body spin).

Usage: Desired tilt $\theta_{cmd} \rightarrow \mathbf{v}_{body} \rightarrow \omega_{wheels}$

Sensor Fusion: Complementary Filter

MPU-6050 provides:

- Accelerometer: Tilt estimate (noisy, no drift)
- Gyroscope: Angular rate (clean, but drifts)

Complementary filter:

$$\theta_{fused}[k] = \alpha \cdot (\theta_{fused}[k-1] + \omega_{gyro} \cdot dt) + (1 - \alpha) \cdot \theta_{accel}$$

where:

- $\alpha \approx 0.98$ (high-pass on gyro, low-pass on accel)
- dt : Sample period
- $\theta_{accel} = \text{atan2}(a_y, a_z)$ for roll

Why Complementary Filter?

- Simple, computationally efficient
- Real-time suitable (Kalman filter is better but more complex)
- Sufficient for ballbot application

PID Control Structure

Independent PID for each axis:

$$\tau_i = K_p e + K_i \int e \, dt + K_d \frac{de}{dt}$$

where $e = \theta_{\text{setpoint}} - \theta_{\text{measured}}$

Discrete implementation:

$$P[k] = K_p \cdot e[k]$$

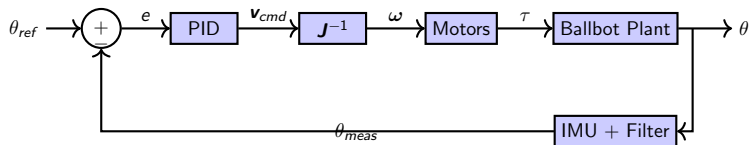
$$I[k] = I[k-1] + K_i \cdot e[k] \cdot dt$$

$$D[k] = K_d \cdot \frac{e[k] - e[k-1]}{dt}$$

$$u[k] = P[k] + I[k] + D[k]$$

Anti-windup: Clamp $I[k]$ to prevent integral saturation

Control Block Diagram



Tilt-Based Locomotion Control

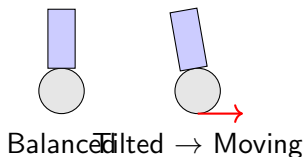
Locomotion via tilt reference:

To move with velocity $\mathbf{v}_{desired}$:

$$\theta_{ref} = K_{locomotion} \cdot \mathbf{v}_{desired}$$

Example:

- Move forward: $\theta_{y,ref} = +0.1$ rad
- Move right: $\theta_{x,ref} = +0.1$ rad
- Stop: $\theta_{ref} = 0$



Control loop: Outer loop (locomotion) sets $\theta_{ref} \rightarrow$ Inner loop (balance) maintains tilt

Control Loop Structure

Main Control Loop:

- 1 Read IMU (I2C)
- 2 Complementary filter
- 3 Compute PID
- 4 Inverse kinematics
- 5 Send step pulses
- 6 Repeat at fixed rate

Timing Requirements:

- Loop frequency: 100-200 Hz
- Jitter: < 1 ms
- Total latency: < 10 ms

ESP32 Task Separation:

Core 0:

- Sensor reading
- Filter update

Core 1:

- Control computation
- Motor commands
- Communication

Uses FreeRTOS for task management

A4988 Current Limiting

Setting motor current via reference voltage:

$$I_{max} = \frac{V_{ref}}{8 \cdot R_{sense}}$$

For A4988 with $R_{sense} = 0.1 \Omega$:

$$I_{max} = \frac{V_{ref}}{0.8}$$

Example calculation:

- NEMA17 rated current: $I_{rated} = 1.5 \text{ A}$
- Set to 70% for margin: $I_{set} = 1.05 \text{ A}$
- Required V_{ref} : $V_{ref} = 0.8 \times 1.05 = 0.84 \text{ V}$

Adjustment: Use small screwdriver on potentiometer while measuring V_{ref} with multimeter

Power Consumption and Battery Life

Power budget:

$$P_{motors} = 3 \times V_{supply} \times I_{avg} = 3 \times 11.1 \times 0.8 = 26.6 \text{ W}$$

$$P_{electronics} \approx 2 \text{ W}$$

$$P_{total} \approx 29 \text{ W}$$

Battery discharge:

For 3S LiPo, 2200 mAh at 11.1V:

$$E_{battery} = 2.2 \times 11.1 = 24.4 \text{ Wh}$$

Estimated runtime:

$$t_{run} = \frac{24.4}{29} \times 0.8 \approx 40 \text{ minutes}$$

(Factor 0.8 accounts for inefficiency and voltage drop)

PID Gain Selection Effects

Gain	Too Low	Too High	Effect
K_p	Slow response	Oscillation	Proportional
K_i	Steady error	Windup, overshoot	Eliminates bias
K_d	Overshoot	Noise amplification	Damping

Tuning procedure:

- 1 Start with all gains at zero
- 2 Increase K_p until small oscillations appear
- 3 Add K_d to dampen oscillations
- 4 Add small K_i to eliminate steady-state error
- 5 Fine-tune iteratively

Typical values (normalized):

$$K_p \approx 10-50, \quad K_i \approx 0.1-1, \quad K_d \approx 1-5$$

Oscillation Analysis

Types of oscillation:

1. High-frequency (> 10 Hz):

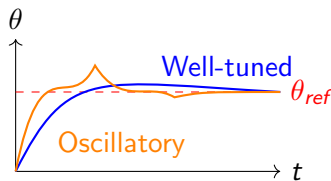
- Cause: Excessive K_d
- Fix: Reduce K_d , add filter

2. Medium frequency (2-5 Hz):

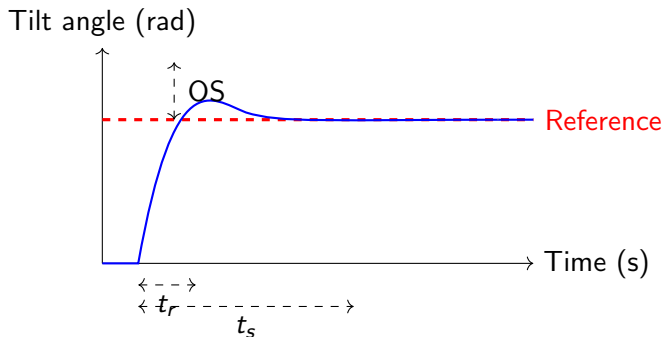
- Cause: Excessive K_p
- Fix: Reduce K_p , increase K_d

3. Low frequency (< 1 Hz):

- Cause: Excessive K_i
- Fix: Reduce K_i , add anti-windup



Step Response Characteristics



Performance metrics:

- Rise time (t_r): Time to reach 90% of reference
- Overshoot (OS): Peak value beyond reference
- Settling time (t_s): Time to stay within $\pm 2\%$ of reference
- Steady-state error: Final error at $t \rightarrow \infty$

Current System Limitations

Modeling Simplifications

- Small angle approximation breaks down at $>15^\circ$
- Neglected friction (ball-roller, bearing, aerodynamic)
- Assumed rigid body (ignores flexibility)
- No slip modeling (critical for rapid maneuvers)

Hardware Constraints

- Stepper motor limitations (torque ripple, finite step size)
- IMU noise and bias drift
- Computational latency in ESP32
- Power constraints (battery weight vs. capacity)

Control Limitations

- PID cannot optimize for multiple objectives

Friction and Slip Modeling

Friction torque at ball-roller interface:

$$\tau_{friction} = \mu N r_{contact}$$

where:

- μ : Coefficient of friction (rubber on ball)
- N : Normal force at contact
- $r_{contact}$: Contact radius

Slip condition:

Slip occurs when commanded torque exceeds maximum static friction:

$$|\tau_{cmd}| > \tau_{friction,max}$$

Effects of slip:

- Loss of position tracking
- Reduced control authority
- Potential instability

Mitigation:

Advanced Control: LQR

Linear Quadratic Regulator (LQR):

Minimize cost function:

$$J = \int_0^{\infty} (\mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{u}^T \mathbf{R} \mathbf{u}) dt$$

where:

- \mathbf{Q} : State cost matrix (penalize deviations)
- \mathbf{R} : Input cost matrix (penalize control effort)

Optimal control law:

$$\mathbf{u}^* = -\mathbf{K} \mathbf{x}$$

where \mathbf{K} is computed by solving the Riccati equation.

Advantages over PID:

- Systematic design (no manual tuning)
- Multi-objective optimization
- Provable stability guarantees

State Estimation: Kalman Filter

Extended Kalman Filter (EKF):

Handles sensor noise and bias:

$$\hat{\mathbf{x}}[k|k] = \hat{\mathbf{x}}[k|k-1] + \mathbf{K}[k](\mathbf{z}[k] - \mathbf{h}(\hat{\mathbf{x}}[k|k-1]))$$

where:

- $\hat{\mathbf{x}}$: State estimate
- \mathbf{z} : Sensor measurements
- \mathbf{K} : Kalman gain
- $\mathbf{h}(\cdot)$: Measurement model

Benefits:

- Optimal fusion of accel and gyro
- Bias estimation and correction
- Better than complementary filter
- Can estimate ball position/velocity

Trade-off: Computational complexity

Field-Oriented Control (FOC) for Motors

Upgrading from steppers to BLDC with FOC:

Advantages:

- Smooth torque output
- Higher efficiency
- Better dynamic response
- Torque control mode
- Higher speed capability

Requirements:

- Current sensing
- Rotor position encoder
- FOC algorithm
- Dedicated motor driver
- Higher computational load

FOC enables:

- Direct torque control (better than velocity control)
- Force/impedance control
- Smoother motion

Future Work and Extensions

Immediate improvements:

- Implement EKF
- Add slip detection
- Vibration damping
- Remote control interface
- Data logging

Medium-term:

- Switch to LQR/LQG
- Add vision system
- Autonomous navigation
- Path planning

Advanced features:

- Machine learning for disturbance rejection
- Adaptive control
- Multi-robot coordination
- Human-robot interaction

Hardware upgrades:

- BLDC with FOC
- Better IMU (BMI088)
- Onboard computer (Jetson Nano)
- LiDAR for obstacle avoidance

Summary

Key Takeaways:

- 1 Ballbot is an underactuated, inherently unstable system
- 2 Requires careful mechanical design (120° roller config, high CoM)
- 3 Mathematical modeling enables systematic control design
- 4 Sensor fusion (complementary filter) provides reliable state estimate
- 5 PID control provides balance; tilt commands enable locomotion
- 6 Hardware selection balances performance, cost, complexity
- 7 Many opportunities for advanced control and AI integration

Core Principle

Maintain controlled instability to achieve dynamic stability

References and Resources

Key Papers:

- Lauwers et al., "A Dynamically Stable Single-Wheeled Mobile Robot with Inverse Mouse-Ball Drive" (2006)
- Kumagai and Ochiai, "Development of a Robot Balancing on a Ball" (2010)

Software/Hardware:

- ESP32 Arduino Core: github.com/espressif/arduino-esp32
- MPU6050 Library: github.com/jrowberg/i2cdevlib
- Stepper control: AccelStepper library

Further Learning:

- Slotine & Li, "Applied Nonlinear Control"
- Åström & Murray, "Feedback Systems"
- Lynch & Park, "Modern Robotics"

Thank You!

Questions?

Contact: `cipher0xx@gmail.com`