

# Human Following Robot || Linear MPC

Pseudocode || Daksh Raval

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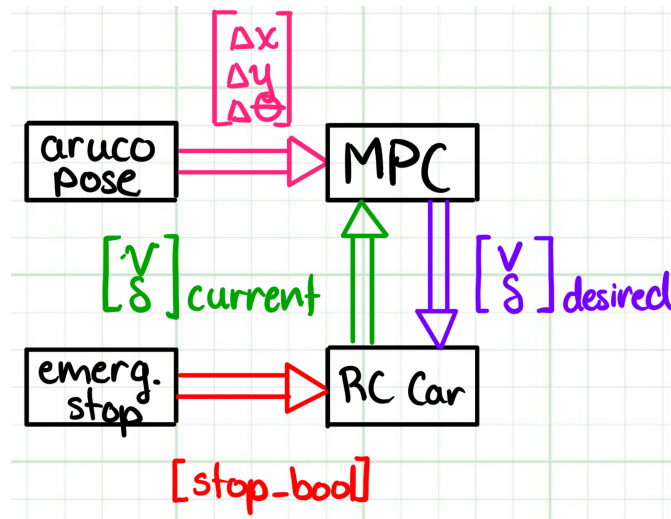
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## 1 ArUco Pose

### 1.1 ROS Node Connections

**Nutshell 1.** One to estimate ArUco pose, other for MPC



**Nutshell 2.** Detect markers, plug into PnP solver, get pose. Mostly out-of-the-box, difficulty lies in formatting chain of data transferring

```
import cv # openCV  
  
MARKER_SIZE = 0.10 # meters
```

```

CAMERA_INTRINSICS = [focal, principal, distortion] # set manually

subscribe(rgb_image)

def main():
    aruco_corners = cv.detectArUco(rgb_image)

    # spits out pose from camera to ArUco in camera/OpenCV frame
    trans_vec, rot_vec = cv.ArUcoPose(aruco_corners, CAMERA_INTRINSICS)

    # camera frame: [Z, Y, X] = [Forward, Down, Right]
    # RC car frame: [X, Y, Z] = [Forward, Left, Up]
    convert_frames(trans_vec, rot_vec)
    != not sure how to convert rotation vector

    # invert so ArUco reference
    pose = - [converted_trans_vec, converted_rot_vec]
    != not sure if i should do this cause camera to ArUco is ArUco - camera

    publish(aruco_pose)

```

## 2 MPC

**Nutshell 3.** *Linearize and discretize at operating point, plug into MPC solver, get trajectory.*

### 2.1 Overarching Pseudocode

**Nutshell 4.** *Functional programming. Callbacks = 'don't bother until you got mail for me'.*

- `make_mpc_template()` : compile boilerplate 'optimization-problem'-object for CasADi solver
- `parse_prev_control_callback()` : unpack actual throttle/steer motor encoder values from `Float32MultiArray`
- `trajectory_planner_callback()` : when new human position comes, compute new trajectory
- `get_linearized_matrices()` : linearize matrices at operating point
- `solve_mpc()` : plug into CasADi's IPOPT solver, get trajectory
- `pack_publish_control()` : turns numpy array into `Twist` and publishes to car
- `differential_wheel_speed()` : computes rear wheel speeds using Ackerman kinematic model

```

def main():
    solver_template = make_mpc_template()

    subscribe('/current_vel_steer', parse_prev_control_callback)
    subscribe('/aruco_pose_topic', trajectory_planner_callback)

def make_mpc_template():
    mpc_template = state_op_param, A_d_param, B_d_param, ..., x_ref_param
    return mpc_template

def parse_prev_control_callback():
    return control_op

def get_linearized_matrices(state_op, control_op, dt, L):
    linearized_matrices = A_d, B_d, d_d
    return linearized_matrices

def trajectory_planner_callback(aruco_pose):

    trajectory = solve_mpc(mpc_template, linearized_matrices, state_op,
        ↪ control_op)
    next_action = trajectory[0]
    pack_publish_control(next_action)
    return next_action

def pack_publish_control(next_action):
    left_wheel_speed, right_wheel_speed = differential_wheel_speed()
    return next_action_Twist_vector

def differential_wheel_speed(velocity, steer):
    return left_wheel_speed, right_wheel_speed

```

## 2.2 Overarching Pseudocode

```

subscribe(aruco_pose) # [x, y, \theta] from aruco tracking/pose estimation
subscribe(speed_steer) # [v, \delta] from car's motor encoders

def parse_state_callback(aruco_pose):

def get_linearized_matrices(state: list, control: list):

horizon = 10 # intervals

```

```

optim = casadi.Optim()

X = optim.variable(3, horizon + 1) # 3 states, 3x11 matrix for storage
U = optim.variable(2, N)

A_d = optim.parameter(3, 3) # matrix size, custom casadi storage object
B_d = optim.parameter(3, 2)
X_current = optim.parameter(3) # only current state vector

optim.subject_to(X[0] == X_current) # start at current state

for k in range(horizon):
    x_next = X[k+1]
    x_current = X[k]
    u_current = U[k]

    # constrains to state-space model
    optim.subject_to(x_next == A_d @ x_current + B_d @ u_current # matrix-vector
        → multiplication

    # control constraints
    optim.subject_to(-1.0, U[0], 1.0) # velocity
    optim.subject_to(-0.45, U[1], 0.45) # steer

### Cost Function setup

# weights
Q = [10, 10, 1] # tracking error penalizer
R = [1, 1] # control effort penalizer

cost = 0
for k in range(horizon):
    state_error = X[k] # pose, minus some offset, is simply the state error
    control_effort = U[k]
    cost = x^T @ Q @ x + u^T @ R @ u

## Actual Trajectory Computation
casadi.minimize(cost)

# main loop
while true:
    x, y, theta = aruco_pose
    v, delta = speed, steer

    dt = time_since_last_loop()

```

```
A_lin = linearize_at_operating_point(v, theta)
B_lin = linearize_at_operating_point(v, theta, delta)
optim.A_d = I + A_lin * dt # I is identity matrix
optim.B_d = B_lin * dt

try:
    trajectory = optim.solve()

    u_next = trajectory[0]
```