Drone Applications, Components and Assembly

Lab 6

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20BRS1193

Aim:

To demonstrate RTL (Return to Launch) capabilities in drones, Model Predictive Control (MPC) utilizing the CasADi library, and establishing a different height for each drone waypoint.

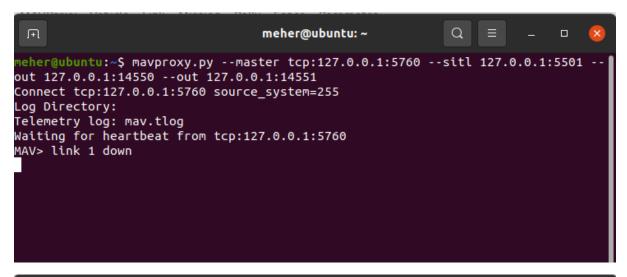
Procedure:

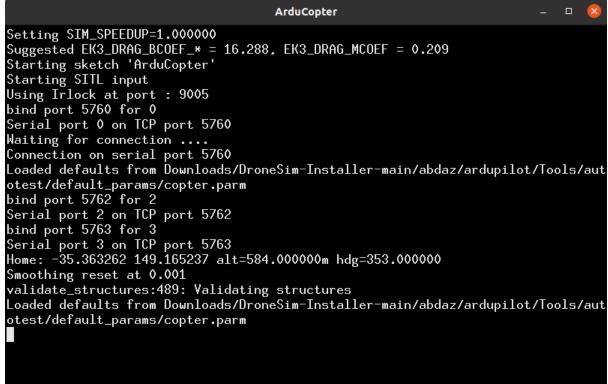
- 1. Install required SITL programs ArduPilot-SITL
- 2. Install mayproxy to connect with SITL
- 3. Install the dronekit Python package
- 4. In a terminal, run the SITL startup command for a copter and in another terminal, run mavproxy:

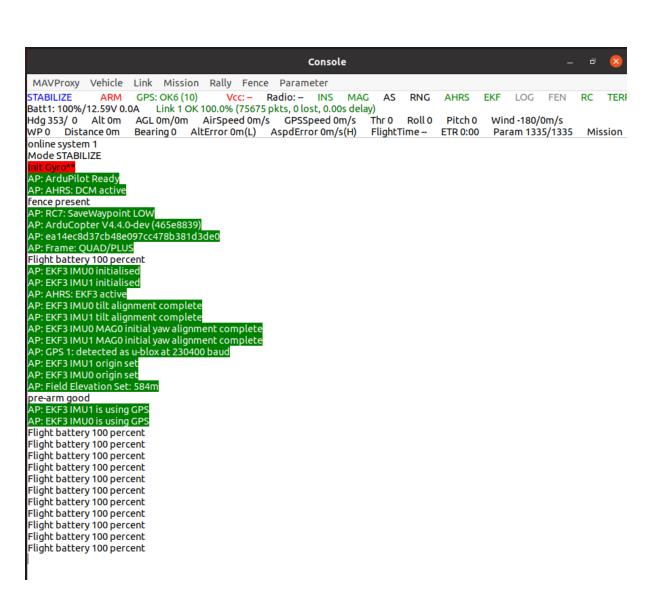
python sim_vehicle.py --map --console -v ArduCopter

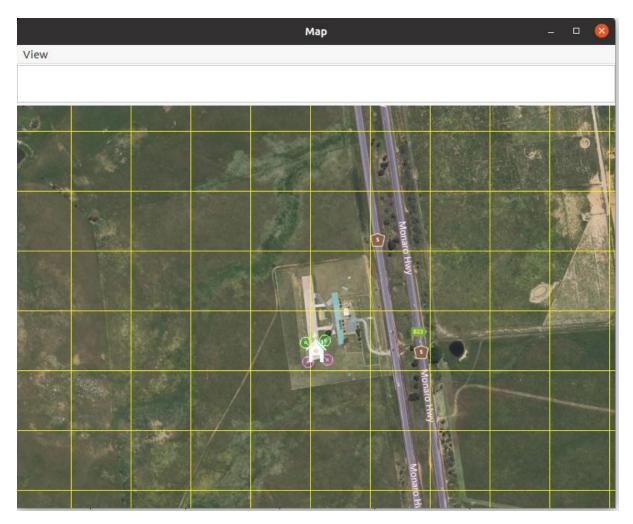
mavproxy.py --master tcp:127.0.0.1:5760 --sitl 127.0.0.1:5501 --out 127.0.0.1:14550 --out 127.0.0.1:14551

```
Ħ
                                       meher@ubuntu: ~
meher@ubuntu:~$ sim_vehicle.py --console --map -v ArduCopter
SIM VEHICLE: Start
SIM_VEHICLE: Killing tasks
SIM_VEHICLE: Starting up at SITL location
SIM VEHICLE: WAF build
SIM VEHICLE: Configure waf
SIM_VEHICLE: "/home/meher/Downloads/DroneSim-Installer-main/abdaz/ardupilot/modules/waf-light" "configure" "--board" "sitl"
Setting top to
Setting out to
Autoconfiguration
                                                : enabled
Setting board to
Using toolchain
Checking for 'g++' (C++ compiler)
Checking for 'gcc' (C compiler)
Checking for c flags '-MMD'
                                                : /usr/lib/ccache/g++
Checking for cxx flags '-MMD'
CXX Compiler
Checking for need to link with librt
                                                : not necessary
Checking for feenableexcept
Enabled OpenDroneID
Enabled firmware ID checking
```









5. In a third terminal, run the required python file. Below are the codes, outputs and terminals of the three python files.

MPC.py

Python code demonstrate Model Predictive Control (MPC) using CasADi library:

import casadi as cs

import numpy as np

import matplotlib.pyplot as plt

Define the system dynamics

A = np.array([[1, 1], [0, 1]])

B = np.array([[0], [1]])

C = np.array([[1, 0], [0, 1]])

D = np.array([[0], [0]])

```
# Define the MPC parameters
N = 5
dt = 0.1
Q = np.diag([1, 1])
R = np.array([[1]])
# Define the optimization problem
opti = cs.Opti()
# Define the state variables
x = opti.variable(2, N+1)
x0 = opti.parameter(2, 1)
# Define the control variables
u = opti.variable(1, N)
# Define the reference trajectory
x ref = opti.parameter(2, N+1)
u ref = opti.parameter(1, N)
# Define the initial state constraint
opti.subject_to(x[:, 0] == x0)
# Define the dynamic constraints
for k in range(N):
  x_next = cs.mtimes(A, x[:, k]) + cs.mtimes(B, u[:, k])
  opti.subject_to(x[:, k+1] == x_next)
```

```
# Define the cost function
J = 0
for k in range(N):
  J += cs.mtimes([(x[:, k] - x_ref[:, k]).T, Q, (x[:, k] - x_ref[:, k])])
  J += cs.mtimes([(u[:, k] - u_ref[:, k]).T, R, (u[:, k] - u_ref[:, k])])
opti.minimize(J)
# Define the control constraints
opti.subject_to(u <= 1)
opti.subject to(u \ge -1)
# Set the initial state parameter
x0_val = np.array([[0], [0]])
opti.set value(x0, x0 val)
# Define the reference trajectory and control inputs
x_ref_val = np.zeros((2, N+1))
x_ref_val[0, :] = np.linspace(0, 1, N+1)
u ref val = np.zeros((1, N))
opti.set value(x ref, x ref val)
opti.set_value(u_ref, u_ref_val)
# Simulate the system and plot the results
x_val = np.zeros((2, N+1))
u_val = np.zeros((1, N))
for i in range(N):
  # Update the optimization problem with the current state
  opti.set_initial(u, u_val)
```

```
opti.set_initial(x, x_val)
  # Solve the optimization problem
  sol = opti.solve()
  # Extract the control input
  u_val = opti.value(u[:, 0])
  # Update the system state
  x_{val}[:, i+1] = np.squeeze(cs.mtimes(A, x_{val}[:, i]) + cs.mtimes(B, u_{val}))
# Plot the results
plt.plot(x_ref_val[0, :], x_ref_val[1, :], 'r--', label='Reference')
plt.plot(x_val[0, :], x_val[1, :], 'b', label='MPC')
plt.legend()
plt.xlabel('x1')
plt.ylabel('x2')
plt.show()
```

```
eher@ubuntu:~/Downloads$ python MPC.py
*****************************
This program contains Ipopt, a library for large-scale nonlinear optimization. Ipopt is released as open source code under the Eclipse Public License (EPL).
           For more information visit https://github.com/coin-or/Ipopt
This is Ipopt version 3.14.11, running with linear solver MUMPS 5.4.1.
Number of nonzeros in equality constraint Jacobian...:
Number of nonzeros in inequality constraint Jacobian.:
                                                                               10
Number of nonzeros in Lagrangian Hessian....:
                                                                               20
Total number of variables..... variables with only lower bounds:
                                                                               17
                                                                                0
                     variables with lower and upper bounds:
                                                                                0
                           variables with only upper bounds:
                                                                                0
Total number of equality constraints......

Total number of inequality constraints......
                                                                               12
                                                                               10
    inequality constraints with only lower bounds: inequality constraints with lower and upper bounds:
                                                                                5
                                                                                Θ
          inequality constraints with only upper bounds:
iter
                           inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
          objective
    0 1.200000e+00 0.00e+00 7.37e-01 -1.0 0.00e+00 1 3.3091670e-01 1.11e-16 1.93e-01 -1.7 7.22e-01 2 3.2274534e-01 1.11e-16 2.00e-07 -1.7 3.68e-02 3 3.2269431e-01 1.11e-16 1.50e-09 -3.8 3.14e-03 4 3.2269430e-01 2.78e-17 1.84e-11 -5.7 4.15e-05 5 3.2269430e-01 5.55e-17 2.51e-14 -8.6 9.66e-08
                                                                       - 0.00e+00 0.00e+00
                                                                           8.37e-01 1.00e+00f
                                                                        - 1.00e+00 1.00e+00f
                                                                        - 1.00e+00 1.00e+00f
                                                                       - 1.00e+00 1.00e+00h
- 1.00e+00 1.00e+00h
Number of Iterations....: 5
                                                                              (unscaled)
                                             (scaled)
Objective.....

Dual infeasibility....:
                                                                      3.2269430051813469e-01
                                    3.2269430051813469e-01
                                    2.5059035640133008e-14
                                                                      2.5059035640133008e-14
                                    5.5511151231257827e-17
Constraint violation...:
                                                                      5.5511151231257827e-17
                                    0.0000000000000000e+00
Variable bound violation:
                                                                      0.0000000000000000e+00
                                    2.5061070880374509e-09
Complementarity....:
                                                                      2.5061070880374509e-09
Overall NLP error...:
                                    2.5061070880374509e-09
                                                                      2.5061070880374509e-09
```

```
Q =
                                                 meher@ubuntu: ~/Downloads
Number of objective function evaluations
Number of objective gradient evaluations
                                                                     = 6
Number of equality constraint evaluations
Number of inequality constraint evaluations
                                                                     = 6
                                                                     = 6
Number of equality constraint Jacobian evaluations = 6
Number of inequality constraint Jacobian evaluations = 6
Number of Lagrangian Hessian evaluations
                                                                     = 5
Total seconds in IPOPT
                                                                     = 0.022
EXIT: Optimal Solution Found.
                                      (avg)
        solver
                  : t_proc
                                                t_wall
                                                                 (avg)
                                                                             n_eval
                     108.00us ( 18.00us) 98.41us ( 16.40us)
         nlp_f
                                                                                  6
         nlp_g | 66.00us ( 11.00us) 63.03us ( 10.50us)
grad_f | 101.00us ( 14.43us) 99.46us ( 14.21us)
ness_l | 90.00us ( 18.00us) 72.69us ( 14.54us)
jac_g | 120.00us ( 17.14us) 111.74us ( 15.96us)
total | 14.17ms ( 14.17ms) 32.66ms ( 32.66ms)
                                                                                   6
  nlp_grad_f
nlp_hess_l
                                                                                   5
    nlp_jac_g
This is Ipopt version 3.14.11, running with linear solver MUMPS 5.4.1.
Number of nonzeros in equality constraint Jacobian...:
Number of nonzeros in inequality constraint Jacobian.:
                                                                                10
Number of nonzeros in Lagrangian Hessian....:
                                                                                20
Total number of variables....:
                                                                                17
                     variables with only lower bounds: variables with lower and upper bounds:
                                                                                0
                                                                                 0
                          variables with only upper bounds:
                                                                                0
Total number of equality constraints....:
                                                                                12
Total number of inequality constraints....:
                                                                                10
          inequality constraints with only lower bounds:
                                                                                5
    inequality constraints with lower and upper bounds:
          inequality constraints with only upper bounds:
                                                                                 5
iter
                                      inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
                           inf_pr
          obiective
      1.5290891e+00 4.68e-01 7.02e-01 -1.0 0.00e+00 - 0.00e+00 0.00e+00
    1 3.7148939e-01 5.55e-17 2.10e-01 -1.0 8.52e-01
2 3.2324261e-01 1.11e-16 2.00e-07 -1.7 2.14e-01
3 3.2269470e-01 1.11e-16 2.83e-08 -2.5 2.70e-02
                                                                        - 8.25e-01 1.00e+00f
    2 3.2324261e-01 1.11e-16 2.00e-07 -1.7 2.14e-01 - 1.00e+00 1.00e+00f

3 3.2269470e-01 1.11e-16 2.83e-08 -2.5 2.70e-02 - 1.00e+00 1.00e+00f

4 3.2269430e-01 1.11e-16 1.50e-09 -3.8 6.59e-04 - 1.00e+00 1.00e+00f

5 3.2269430e-01 1.11e-16 1.84e-11 -5.7 9.17e-06 - 1.00e+00 1.00e+00h

6 3.2269430e-01 1.11e-16 2.51e-14 -8.6 9.54e-08 - 1.00e+00 1.00e+00h
Number of Iterations....: 6
                                              (scaled)
                                                                              (unscaled)
Objective....:
                                   3.2269430051813475e-01
                                                                      3.2269430051813475e-01
Dual infeasibility....:
                                                                      2.5059035640133008e-14
                                   2.5059035640133008e-14
Constraint violation...:
                                    1.1102230246251565e-16
                                                                      1.1102230246251565e-16
Variable bound violation:
                                   0.0000000000000000e+00
                                                                      0.00000000000000000e+00
```

RTL.py

Python code to simulate RTL (Return to Launch) functionality in drones:

import time

from dronekit import connect, VehicleMode, LocationGlobalRelative import math

```
def get_distance_metres(aLocation1, aLocation2):
  dlat = aLocation2.lat - aLocation1.lat
  dlong = aLocation2.lon - aLocation1.lonreturn math.sqrt((dlat*dlat) + (dlong*dlong)) *
1.113195e5
def distance_to_current_waypoint(awp):
  distancetopoint = get distance metres(vehicle.location.global frame, awp)
  return distancetopoint
# Connect to the PX4 vehicle
#connection_string = 'udp:127.0.0.1:14550'
#vehicle = connect(connection string, wait ready=True)
vehicle = connect('udp:127.0.0.1:14550')
print("Connected!")
# Set the vehicle mode to GUIDED
vehicle.mode = VehicleMode("GUIDED")
# Arm the vehicle
vehicle.armed = True
while not vehicle.armed:
  print("Waiting for vehicle to arm...")
  time.sleep(1)
vehicle.simple takeoff(10)
# Wait for the drone to reach a certain altitude
while True:
  altitude = vehicle.location.global_relative_frame.alt
  if altitude >= 9.5:
    break
```

```
time.sleep(1)
# Define the mission waypoints
waypoints = [
  LocationGlobalRelative(-35.363261, 149.165230, 10),
  LocationGlobalRelative(-35.362933, 149.164652, 10),
  LocationGlobalRelative(-35.363275, 149.164340, 10),
  LocationGlobalRelative(-35.363700, 149.164889, 10)
1
# Move to each waypoint in turn with a fixed altitude of 20 meters
for waypoint in waypoints:
  # Set the target waypoint with a fixed altitude of 20 meters
  target altitude = 20
  target location = LocationGlobalRelative(
    waypoint.lat, waypoint.lon, target_altitude)
  vehicle.simple_goto(target_location)
  # Wait for the vehicle to reach the waypoint
  while True:
    current_pos = vehicle.location.global_relative_frame
    dist = current_pos.distance_to(target_location)
    if dist < 1:
      break
    time.sleep(1)
# Set the vehicle mode to RTL (Return to Launch)
vehicle.mode = VehicleMode("RTL")
```

Wait for the vehicle to return to the launch point and land while vehicle.armed:

print("Waiting for vehicle to land...")

time.sleep(1)

Disconnect from the vehicle

vehicle.close()

meher@ubuntu:~/Downloads\$ python RTL.py
Connected!
Waiting for vehicle to arm...





```
Connected!
Waiting for vehicle to arm...
Waiting for vehicle to land...
```

WAYP.py

Python code to simulate varying altitude of waypoints in drones:

import time

from dronekit import connect, VehicleMode, LocationGlobalRelative import math

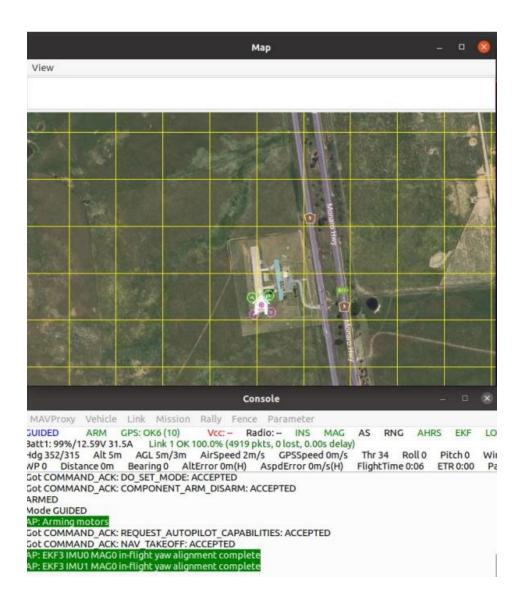
```
def get_distance_metres(aLocation1, aLocation2):
    dlat = aLocation2.lat - aLocation1.lat
    dlong = aLocation2.lon - aLocation1.lonreturn math.sqrt((dlat*dlat) + (dlong*dlong)) * 1.113195e5

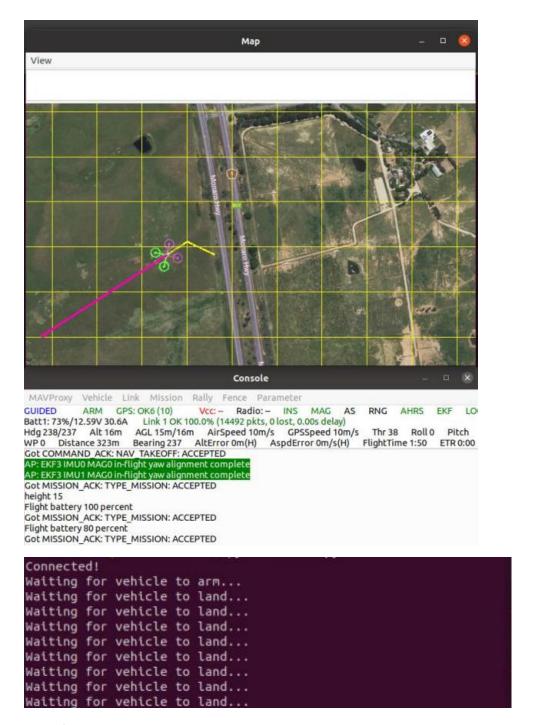
def distance_to_current_waypoint(awp):
    distancetopoint = get_distance_metres(vehicle.location.global_frame, awp)
    return distancetopoint
```

```
# Connect to the PX4 vehicle
#connection_string = 'udp:127.0.0.1:14550'
#vehicle = connect(connection_string, wait_ready=True)
vehicle = connect('udp:127.0.0.1:14550')
print("Connected!")
# Set the vehicle mode to GUIDED
vehicle.mode = VehicleMode("GUIDED")
# Arm the vehicle
vehicle.armed = True
while not vehicle.armed:
  print("Waiting for vehicle to arm...")
  time.sleep(1)
vehicle.simple_takeoff(10)
# Wait for the drone to reach a certain altitude
while True:
  altitude = vehicle.location.global_relative_frame.alt
  if altitude >= 9.5:
    break
  time.sleep(1)
# Define the mission waypoints
waypoints = [
  LocationGlobalRelative(-35.363261, 149.165230, 10),
  LocationGlobalRelative(-35.362933, 149.164652, 10),
  LocationGlobalRelative(-35.363275, 149.164340, 10),
  LocationGlobalRelative(-35.363700, 149.164889, 10)
]
```

```
# Set the target waypoint with a varying altitude
target_altitude = waypoints.index(waypoint) * 5 + 10
target_location = LocationGlobalRelative(waypoint.lat, waypoint.lon, target_altitude)
vehicle.simple_goto(target_location)
# Wait for the vehicle to reach the waypoint
while True:
  current_pos = vehicle.location.global_relative_frame
  dist = current_pos.distance_to(target_location)
  if dist < 1:
    break
  time.sleep(1)
# Set the vehicle mode to LAND
vehicle.mode = VehicleMode("LAND")
# Wait for the vehicle to land
while vehicle.armed:
  print("Waiting for vehicle to land...")
  time.sleep(1)
# Disconnect from the vehicle
vehicle.close()
```

meher@ubuntu:~/Downloads\$ python WAYP.py
Connected!





Conclusion:

We develop and execute the corresponding Python programmes to demonstrate Model Predictive Control (MPC) with the CasADi library, RTL (Return to Launch) functionality in drones, and setting a variable height for each waypoint of a drone.