Q1. Kalman Filter with GPS

With linear position measurement model

(b) Plot true traj and velocity

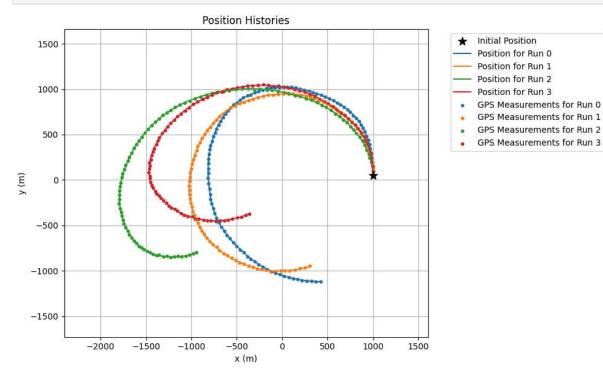
```
In []: # (b)
    import numpy as np
    import matplotlib.pyplot as plt
    import q1q2_simulator_class as sim

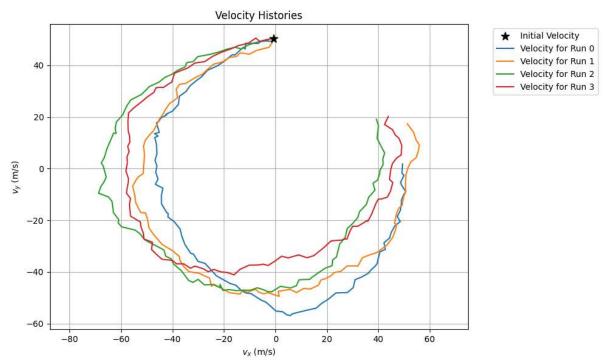
# Define initial state [position p, velocity s]
    n_steps = 100
    p0 = [1000, 0]
    s0 = [0, 50]

# Initialize the simulator
    quadrotor = sim.QuadrotorSimulator(sensor="GPS") # defualt noise

# Simulate result for multi-runs
    pos_hists, vel_hists, meas_pose_hists, uhists = quadrotor.simulate_multiple_runs(p@

# Plot simulated traj and velocity
    fig1_b_pos = quadrotor.plot_position_histories(pos_hists, meas_pose_hists)
    fig1_b_vel = quadrotor.plot_velocity_histories(vel_hists)
```





(c) KF - Position Measurement

```
In [ ]: mu_0 = np.array([1500, 100, 0, 55]) # 4x1
                       sigma_0 = np.block([[250000*np.eye(2), np.zeros((2,2))], [np.zeros((2,2)), np.eye(2), np.eye(2), np.eye(2), np.eye(3), np.eye(4), np.eye(4), np.eye(4), np.eye(4), np.eye(5), np.eye(6), 
                       # Define matrix
                       dt = quadrotor.dt
                       A = np.block([[np.eye(2), dt * np.eye(2)],
                                                       [np.zeros((2,2)), np.eye(2)]]) # (4x4) state transition matrix
                       B = np.block([[np.zeros((2,2))], [dt*np.eye(2)]]) # (4x2) control-input matrix
                       C = np.block([np.eye(2), np.zeros((2,2))]) # (4x2)
                       # Print matrices
                       print("State Transition Matrix (A):")
                       print(A)
                       print("\nControl-Input Matrix (B):")
                       print(B)
                       print("\nObservation Matrix (C):")
                       print(C)
                       # Define noise matrix
                       Q = np.block([[np.zeros((2,2)), np.zeros((2,2))], [np.zeros((2,2)), quadrotor.q_mat])
                       R = quadrotor.r_mat
                       def KF(mu_0, sigma_0, u_arr, y_arr, A, B, C, Q, R):
                                  # Place Holder
                                 mu_update = [mu_0] # udpate state mean
                                  cov_update = [sigma_0] # udpate state cov
                                 mu_pred = [] # pred state mean
                                 mu_cov = [] # pred state cov
                                 for t, (u, y) in enumerate(zip(u_arr, y_arr)):
                                            # Prediction Step
                                            # print(u_arr.shape, u.shape)
                                            mu_bar_next = A @ mu_update[-1] + B @ u # Predict the next state, 4x1
                                            sigma_bar_next = A @ cov_update[-1] @ A.T + Q # Predict the next covariand
                                            # Update Step
                                            K_t_numerator = sigma_bar_next @ C.T # (4x4)(4x2) = (4x2)
```

```
K_t_denominator = C @ sigma_bar_next @ C.T + R # 2x2
                 K_t = K_t_numerator @ np.linalg.inv(K_t_denominator) # 4x2
                 # observation
                expected y = C @ mu bar next # pass the expected state through the measurem
                mu next = mu bar next + K t @ (y - expected y) # Update state mean
                 sigma_next = (np.eye(4) - K_t @ C) @ sigma_bar_next # Update state covaria
                mu update.append(mu next)
                 cov_update.append(sigma_next)
                mu_pred.append(mu_bar_next)
                mu cov.append(sigma bar next)
             return np.array(mu_update[1:]), np.array(cov_update[1:]), np.array(mu_pred), np.array
        State Transition Matrix (A):
        [[1. 0. 1. 0.]
         [0. 1. 0. 1.]
         [0. 0. 1. 0.]
         [0. 0. 0. 1.]]
        Control-Input Matrix (B):
        [[0. 0.]
         [0. 0.]
         [1. 0.]
         [0. 1.]]
        Observation Matrix (C):
        [[1. 0. 0. 0.]
         [0. 1. 0. 0.]]
In [ ]: | # plot error ellipse from Pset2
        import scipy
        from scipy.stats import chi2
         def error_ellipse(ax, mu, sig, p=0.95, samples=1000):
            # creat ellipse
            t = np.linspace(0, 2*np.pi, 100)
            circ = np.array([np.cos(t), np.sin(t)]).T
            r = chi2.ppf(p, df=2)
            eta = scipy.linalg.sqrtm(sig) * np.sqrt(r)
            ellipse = circ @ eta + mu
             # Draw
             ax.plot(ellipse[:,0], ellipse[:,1], color='k', alpha=0.2, label="Error")
```

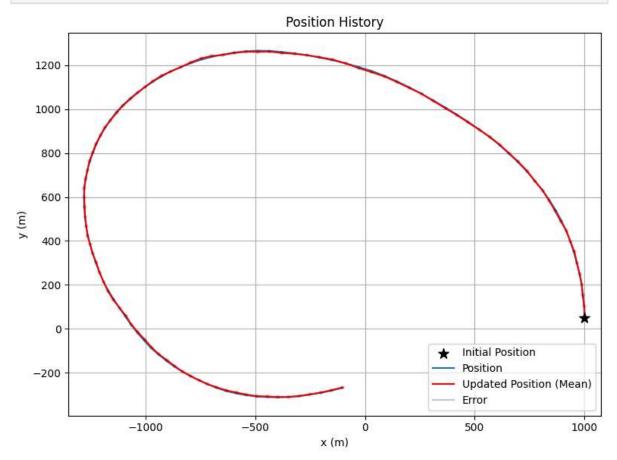
Plot KF (positon)

```
In []: # Simulate result for single run
pos_hist, vel_hist, meas_pose_hist, uhist = quadrotor.simulate(p0=p0, s0=s0, num_t=
# Calculate updated and predicted state (mean and cov) using KF
mu_update, cov_update, mu_pred, mu_cov = KF(mu_0=mu_0, sigma_0=sigma_0, u_arr=uhist

# PLOT result (position)
fig1_c_pos = quadrotor.plot_position_history(pos_hist, show_plot=False) # plot true
ax = fig1_c_pos.axes[0]
ax.plot(mu_update[:,0], mu_update[:,1], color='r', label="Updated Position (Mean)")

for t in range(mu_update.shape[0]):
    mean = mu_update[t,:2] # mean for pos
    cov = cov_update[t,:2, :2] # cov for pos
    error_ellipse(ax, mean, cov)
```

```
handles, legends = ax.get_legend_handles_labels()
plt.legend(handles[:4], legends[:4])
plt.show()
```

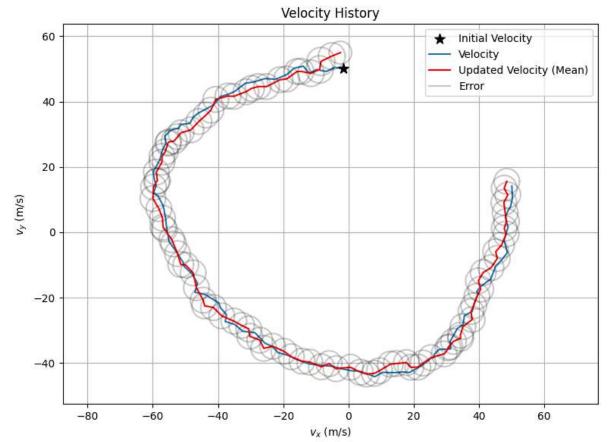


Plot KF (Velocity)

```
In []: # PLOT result (velocity)
fig1_d_vel = quadrotor.plot_velocity_history(vel_hist, show_plot=False) # plot true
ax = fig1_d_vel.axes[0]
ax.plot(mu_update[:,2], mu_update[:,3], color='r', label="Updated Velocity (Mean)")

for t in range(mu_update.shape[0]):
    mean = mu_update[t,2:] # mean for vel
    cov = cov_update[t, 2:, 2:] # cov for vel
    error_ellipse(ax, mean, cov)

handles, legends = ax.get_legend_handles_labels()
plt.legend(handles[:4], legends[:4])
plt.show()
```



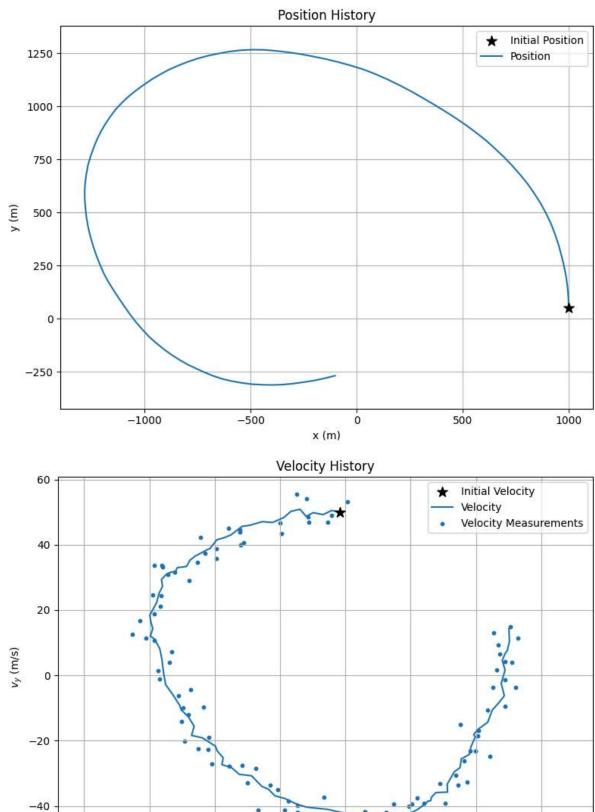
Q2. Kalman Filtering with Inertial Systems

with velocity measurement

```
In []: # Define new state estimation
mu_0_d = np.array([1000, 0, 0, 50])
sigma_0_d = np.block([[np.eye(2), np.zeros((2,2))], [np.zeros((2,2)), np.eye(2)]])
# Define matrix (A, B, Q, R -- unchange)
C = np.block([np.zeros((2,2)), np.eye(2)]) # (4x2)
```

Plot (1) True trajectory (2) true velocity and measurement

```
In [ ]: # Initialization
   quadrotor = sim.QuadrotorSimulator(sensor="Velocity") # Initialize the simulator
   pos_hist, vel_hist, meas_pose_hist, uhist = quadrotor.simulate(p0=p0, s0=s0, num_t=
   fig = quadrotor.plot_position_history(pos_hist, show_plot=True)
   fig = quadrotor.plot_velocity_history(vel_hist, meas_pose_hist, show_plot=True)
```



Plot KF - velocity measurement

-60

-40

```
In []: # Calculate updated and predicted state (mean and cov) using KF
mu_update, cov_update, mu_pred, mu_cov = KF(mu_0=mu_0_d, sigma_0=sigma_0_d, u_arr=u)
# PLOT result (position)
fig2_pos = quadrotor.plot_position_history(pos_hist, show_plot=False) # plot true t
```

-20

20

0 v_x (m/s) 40

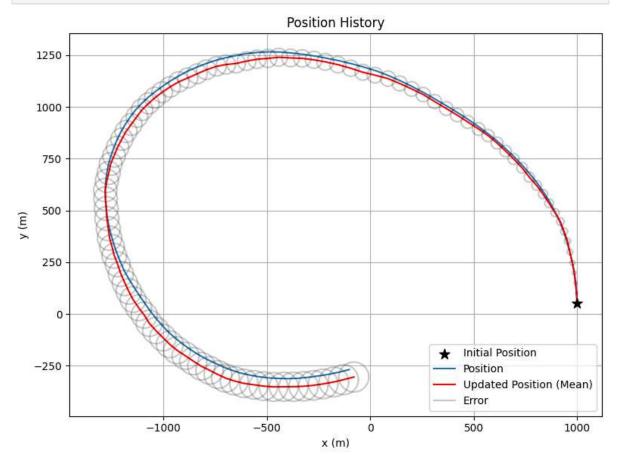
60

-80

```
ax = fig2_pos.axes[0]
ax.plot(mu_update[:,0], mu_update[:,1], color='r', label="Updated Position (Mean)")

for t in range(mu_update.shape[0]):
    mean = mu_update[t,:2] # mean for pos
    cov = cov_update[t,:2, :2] # cov for pos
    error_ellipse(ax, mean, cov)

handles, legends = ax.get_legend_handles_labels()
plt.legend(handles[:4], legends[:4])
plt.show()
```



```
In []: # PLOT result (velocity)
fig2_vel = quadrotor.plot_velocity_history(vel_hist, show_plot=False) # plot true t
ax = fig2_vel.axes[0]
ax.plot(mu_update[:,2], mu_update[:,3], color='r', label="Updated Velocity (Mean)")

for t in range(mu_update.shape[0]):
    mean = mu_update[t,2:] # mean for vel
    cov = cov_update[t, 2:, 2:] # cov for vel
    error_ellipse(ax, mean, cov)

handles, legends = ax.get_legend_handles_labels()
plt.legend(handles[:4], legends[:4])
plt.show()
```

