

Problem 1

Consider an aircraft moving in a plane with constant speed (the magnitude of the velocity vector) and turning with a constant angular rate. Such a model is often used by air traffic control tracking algorithms to describe aircraft executing coordinated turns. Given 2D inertial position variables $\xi(t)$, east position, and $\eta(t)$, north position. The equations of motion are

$$\begin{aligned}\ddot{\xi} &= -\Omega\dot{\eta} \\ \ddot{\eta} &= \Omega\dot{\xi}\end{aligned}$$

where Ω is the constant angular rate, such that $\Omega > 0$ implies a counterclockwise turn. Using the state representation $x(t) = [\xi, \dot{\xi}, \eta, \dot{\eta}]^T$, it can be shown that

$$e^{A\Delta t} = \begin{bmatrix} 1 & \frac{\sin(\Omega\Delta t)}{\Omega} & 0 & -\frac{1-\cos(\Omega\Delta t)}{\Omega} \\ 0 & \cos(\Omega\Delta t) & 0 & -\sin(\Omega\Delta t) \\ 0 & \frac{1-\cos(\Omega\Delta t)}{\Omega} & 1 & \frac{\sin(\Omega\Delta t)}{\Omega} \\ 0 & \sin(\Omega\Delta t) & 0 & \cos(\Omega\Delta t) \end{bmatrix}$$

Where A is the CT LTI state matrix for the system.

part (a)

Find A ignoring measurements y and process noise for now. What is the DT LTI model for this system (also ignoring process noise)?

The A matrix for the continuous time system is:

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & -\Omega \\ 0 & 0 & 0 & 1 \\ 0 & \Omega & 0 & 0 \end{bmatrix}$$

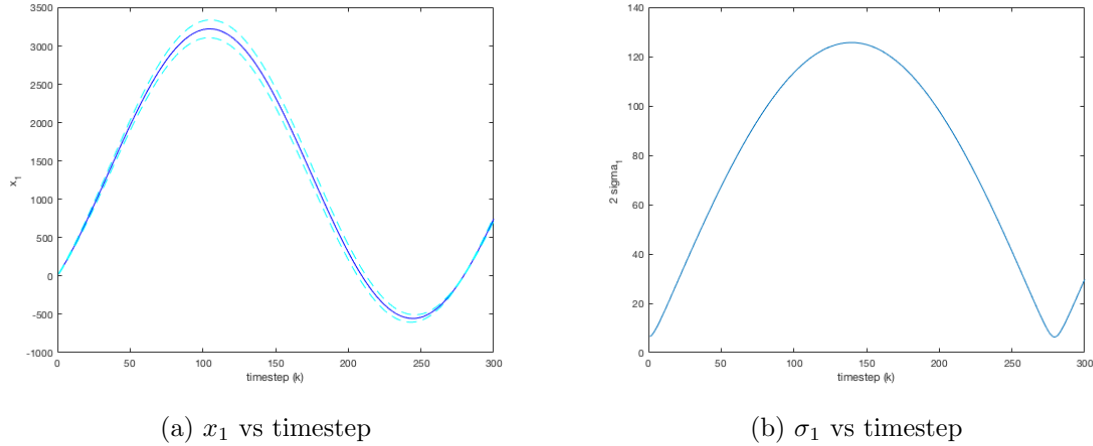
Because there is no control input and no process noise, the discrete time system is described $x(k) = Fx(k-1)$ where $F = \exp\{A\Delta t\}$ as given in the problem statement.

part (b)

For the following questions (i)-(ii) assume a sampling time of $\Delta t = 0.5\text{s}$, $\Omega = 0.045\frac{\text{rad}}{\text{s}}$, and an initial DT state uncertainty $x(0) \sim \mathcal{N}(\mu(0), P(0))$ at $k = 0$. Again, disregard process noise and measurements for now.

part (i)

Derive an expression for the predicted DT state mean $\mu(k) = E[x(k)]$ and state covariance $P(k) = E[(x(k) - \mu(k))(x(k) - \mu(k))^T]$ for an arbitrary number of steps $k = N$ into the future, starting from $k = 0$. Be sure to express your answer in terms of N , $\mu(0)$, and $P(0)$ and the DT state transition matrix F .

Figure 1: Plots for x_1 and σ_1 in meters

Because there is no additive process noise, the state transition from $x(0)$ to $x(1)$ is a linear function of $x(0)$. In fact, any $x(k)$ can be expressed as a function of $x(0)$ per the following:

$$\begin{aligned} x(1) &= Fx(0) \\ x(2) &= Fx(1) = F^2x(0) \\ x(k) &= F^kx(0) \end{aligned}$$

So any $x(k)$ is a linear function of the random variable $x(0)$. This means the mean $\mu(N) = F^N\mu(0)$ and $P(N) = F^N P(0)(F^N)^T$.

part (ii)

Assume

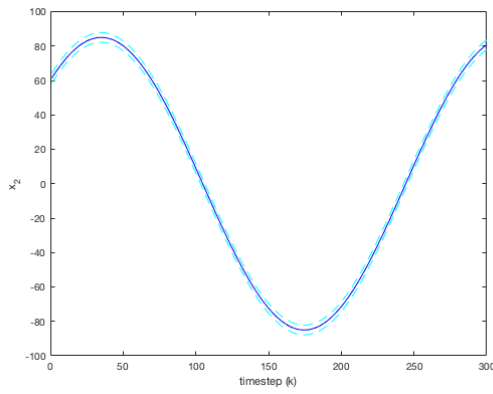
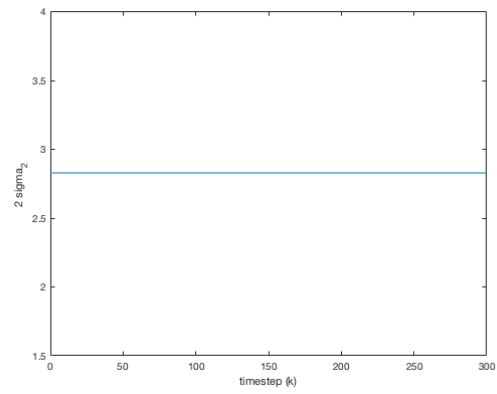
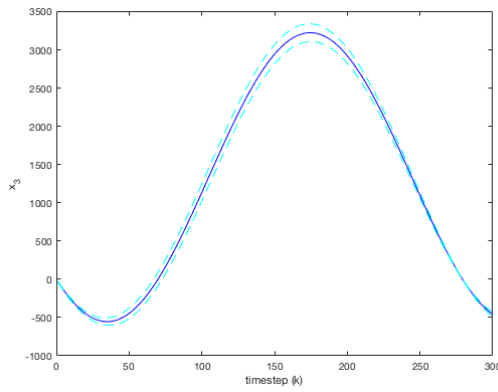
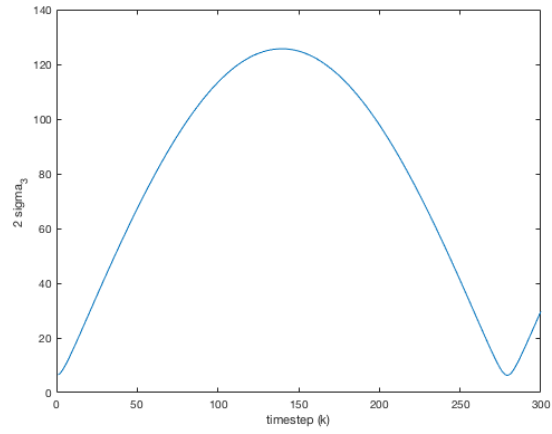
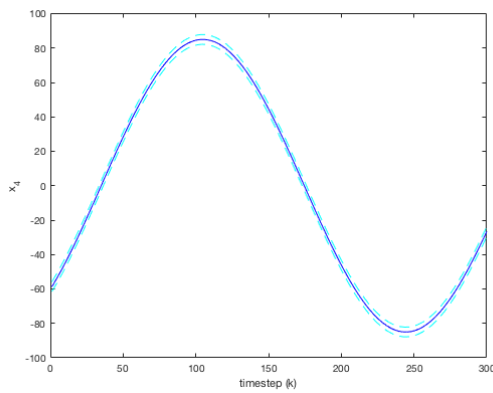
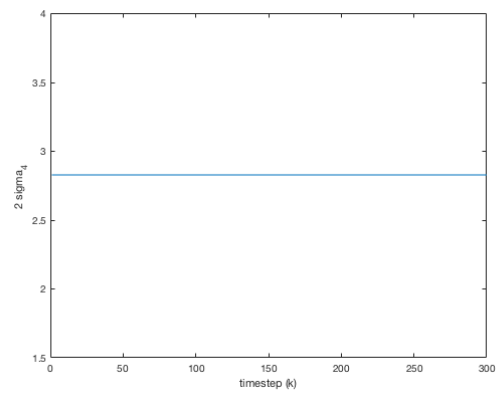
$$\begin{aligned} \mu(0) &= [0\text{m}, 85 \cos(\pi/4)\text{m/s}, 0\text{m}, -85 \sin(\pi/4)\text{m/s}]^T \\ P_a(0) &= \text{diag}([10\text{m}^2, 2(\text{m/s})^2, 10\text{m}^2, 2(\text{m/s})^2]) \end{aligned}$$

Perform a 300 step calculation of $\mu(k)$ and $P(k)$ for $k = 1, 2, \dots, 300$ starting from these initial conditions. Plot the following in separate plots:

- each state element of $\mu(k)$ versus time, along with $\pm 2\sigma$ upper/lower bounds showing the uncertainty for each state at each time k .
- only the positive 2σ values versus time for each state at time k .

Be sure to label all axis with appropriate units and comment on the results.

The plots in figures 1 to 4 show the values of each x variable bounded by 2σ in subplot a and the value for 2σ in subplot b. Note the variance for the position variables x_1 and x_3 appears to be periodic over time but the variance on the velocities x_2 and x_4 is constant.

(a) x_2 vs timestep(b) σ_1 vs timestepFigure 2: Plots for x_2 and σ_2 in m/s(a) x_3 vs timestep(b) σ_3 vs timestepFigure 3: Plots for x_3 and σ_3 in meters(a) x_4 vs timestep(b) σ_4 vs timestepFigure 4: Plots for x_4 and σ_4 in m/s

part (c)

Consider now computing the probability of collision for two aircraft turning independently at the same altitude at the same time, where the initial position and velocity of the aircraft are unknown. Let x_a denote the state of aircraft a and let x_b denote the state of aircraft b . A collision will simply be modeled as an event where both aircraft occupy a bounded rectangular region of airspace at the same time. That is, if $r_a(k) = [\xi_a(k), \eta_a(k)]^T$ and $r_b(k) = [\xi_b(k), \eta_b(k)]^T$, a collision occurs at time k whenever $r_c(k) = r_a(k) - r_b(k)$ lies inside the region R defined by $\Delta\xi(k) \in [-\xi_R, \xi_R]$ and $\Delta\eta(k) \in [-\eta_R, \eta_R]$, where $\Delta\xi(k) = \xi_a(k) - \xi_b(k)$, $\Delta\eta(k) = \eta_a(k) - \eta_b(k)$, and ξ_R and η_R are known constants. At time $k = 0$, suppose both aircraft states are independent and normally distributed, with $x_a(0) \sim \mathcal{N}(\mu_a(0), P_a(0))$ and $x_b(0) \sim \mathcal{N}(\mu_b(0), P_b(0))$. Also assume that $\Delta t = 0.5\text{s}$, $\Omega_a = 0.045\text{rad/s}$ and $\Omega_b = -0.045\text{rad/s}$. Again, ignore process noise and measurements.

part (i)

Derive the expression for the mean and covariance parameters of the pdf $p(r_c(k))$ at any time k , given that $r_c(k) \sim \mathcal{N}(\mu_{r_c}(k), P_{r_c}(k))$ (i.e. the pdf for $r_c(k)$ must be a multivariate Gaussian). Your answers should be in terms of F_a and F_b (the STMs for each aircraft), as well as $\mu_a(0)$, $\mu_b(0)$, and $P_a(0)$, $P_b(0)$. (**Hint:** the concept of marginalizing multivariate Gaussian pdfs will be useful here.)

We can start by finding the distribution for the difference of the full states x_a and x_b .

$$\begin{aligned}\mu_{x_c}(0) &= \mu_{x_a}(0) - \mu_{x_b}(0) \\ \mu_{x_c}(k) &= \mu_{x_a}(k) - \mu_{x_b}(k) \\ &= F_a^k \mu_{x_a}(0) - F_b^k \mu_{x_b}(0) \\ P_{x_c}(0) &= P_{x_a}(0) + P_{x_b}(0) \\ P_{x_c}(k) &= F_a^k P_{x_a}(0) (F_a^k)^T + F_b^k P_{x_b}(0) (F_b^k)^T\end{aligned}$$

Now that we have the distribution for $x_c(k) \sim \mathcal{N}(\mu_{x_c}(k), P_{x_c}(k))$ we can marginalize out the velocity terms to get the distribution for r_c :

$$\begin{aligned}r_c(k) &\sim \mathcal{N}(\mu_c(k), P_c(k)) \\ \mu_c(k) &= J \mu_{x_c}(k) \\ P_c(k) &= J P_{x_c}(k) J^T\end{aligned}$$

where

$$J = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

So the full expression for the mean and covariance of $p(r_c(N))$ is

$$\begin{aligned}\mu_c(N) &= J(F_a^N \mu_{x_a}(0) - F_b^N \mu_{x_b}(0)) \\ P_c(N) &= J(F_a^N P_{x_a}(0) (F_a^N)^T + F_b^N P_{x_b}(0) (F_b^N)^T) J^T\end{aligned}$$

part (ii)

Use the result from part (i) to derive an integral expression for the probability of collision at any time k .

$$p(c(k)) = \int_{-\xi_R}^{\xi_R} \int_{-\eta_R}^{\eta_R} \mathcal{N}(\mu_c(k), P_c(k))$$

part (iii)

Plot the probability of collision vs time using a simulation run of 300 seconds starting from the following initial conditions

$$\begin{aligned}\mu_a(0) &= [0\text{m}, 85 \cos(\pi/4)\text{m/s}, 0\text{m}, -85 \sin(\pi/4)\text{m/s}]^T \\ P_a(0) &= \text{diag}([10\text{m}^2, 4(\text{m/s})^2, 10\text{m}^2, 4(\text{m/s})^2]) \\ \mu_b(0) &= [3200\text{m}, 85 \cos(\pi/4)\text{m/s}, 3200\text{m}, -85 \sin(\pi/4)\text{m/s}]^T \\ P_b(0) &= \text{diag}([11\text{m}^2, 3.5(\text{m/s})^2, 11\text{m}^2, 3.5(\text{m/s})^2])\end{aligned}$$

and using $\xi_R = 100\text{m}$, $\eta_R = 100\text{m}$. Be sure to label axes with appropriate units, and comment on the results. In particular, at what time steps are the vehicles in greatest danger of colliding? Also, what happens to the probability of collision as time increases? What explains this behavior? Matlab's `mvncdf.m` command is useful here.

The plot of the collision probability over timesteps 0 through 300 is shown in figure 5. The probability of collision is greatest at timesteps 80 and 201 which correspond to times 40s and 105s. The probability of collision is generally 0 except at those two times, when the two airplanes are the closest to each other.

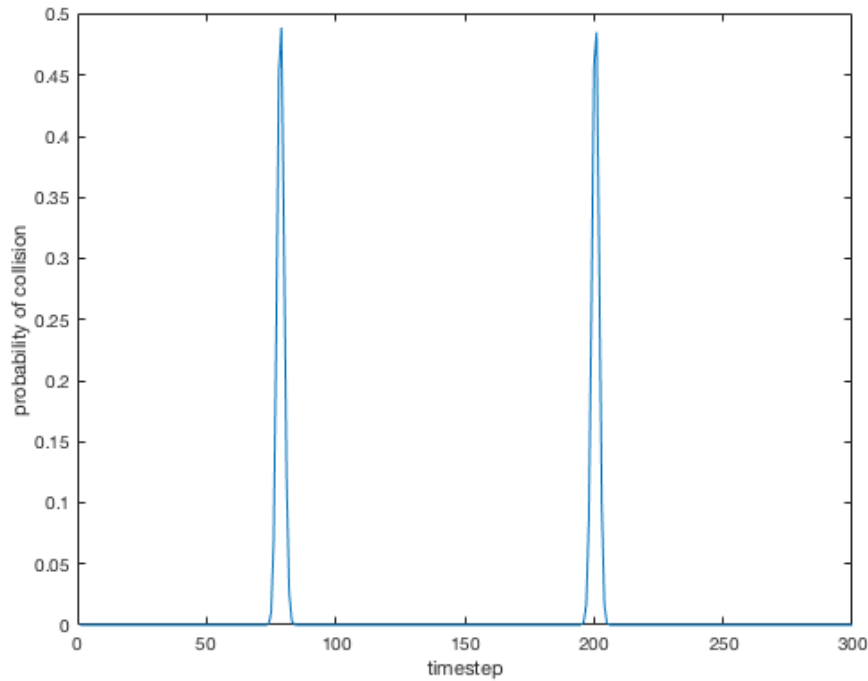


Figure 5: Probability of collision vs timestep